

CORNELL
NYCTECH



deviled eggs
logprob: -1.52



lasagna
logprob: -4.02



spaghetti carbonara
logprob: -3.81



pulled pork sandwiches
logprob: -3.43



french dip sandwiches
logprob: -5.78



egg salad sandwiches
logprob: -6.05



fish tacos
logprob: -0.27



french toast
logprob: -4.55

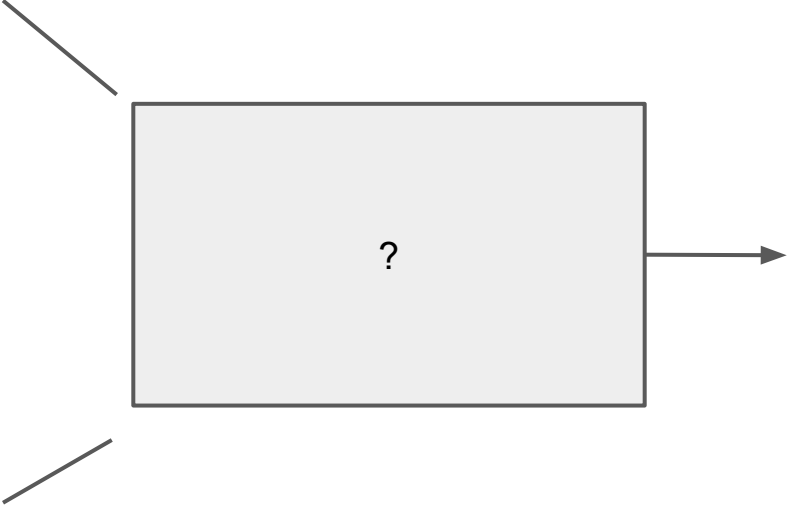


tiramisu
logprob: -5.16

Image Representations and New Domains in Neural Image Captioning

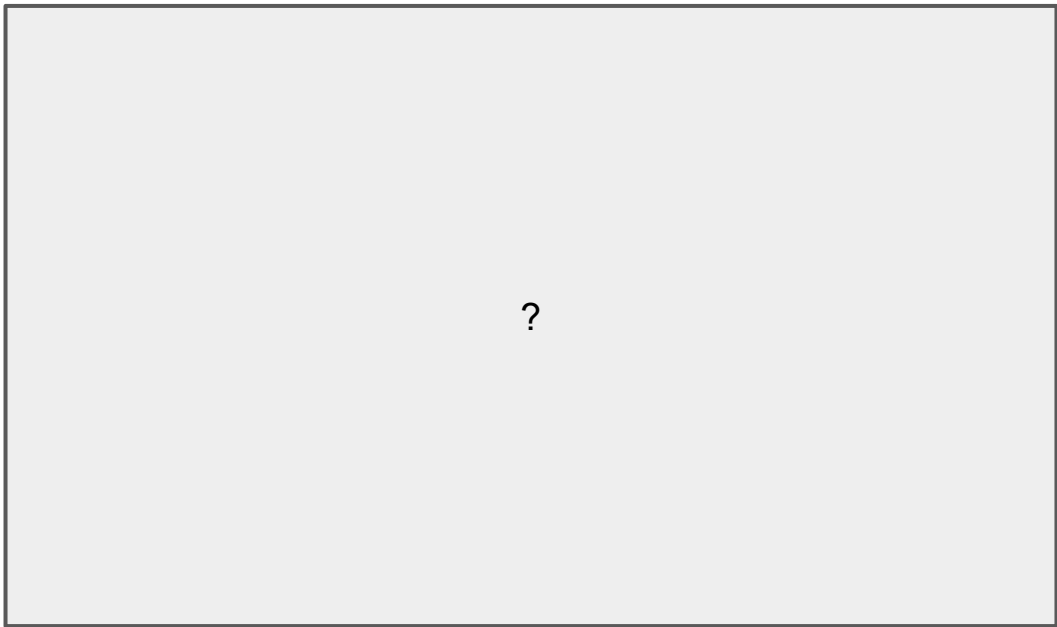
Jack Hessel, Nic Saava, Mike Wilber

The caption generation problem...

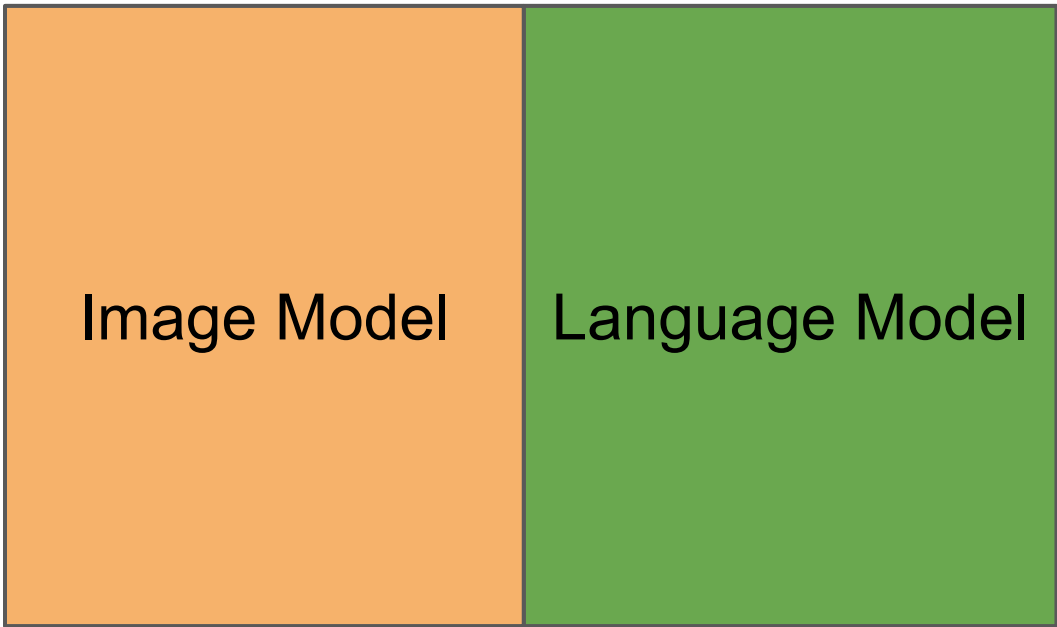


A young girl runs through a grassy field.

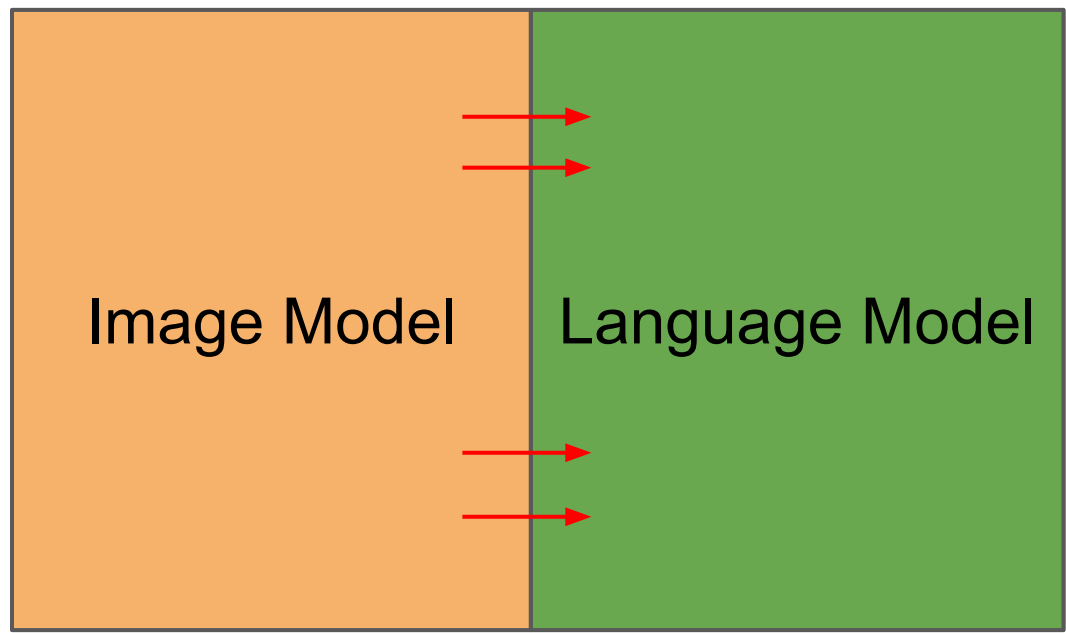
1



A you
through



A you
throug

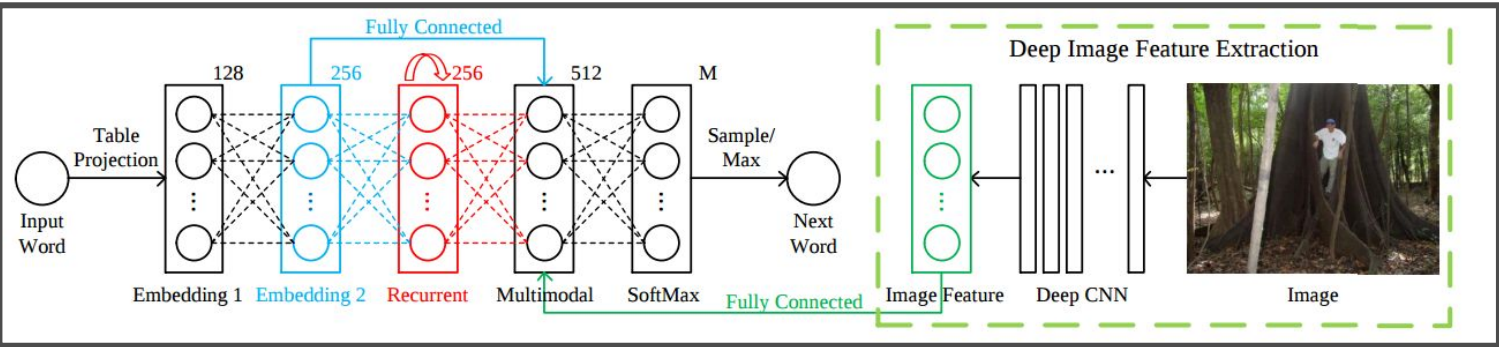
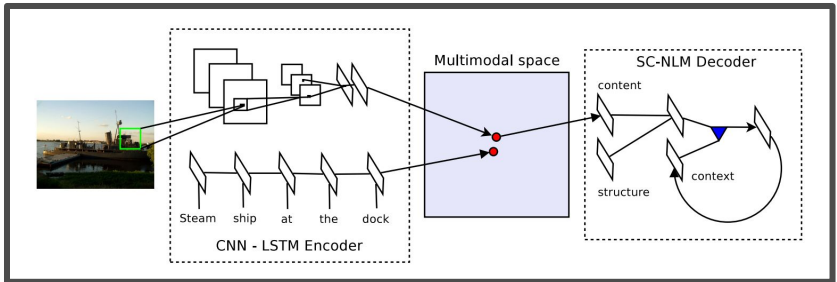
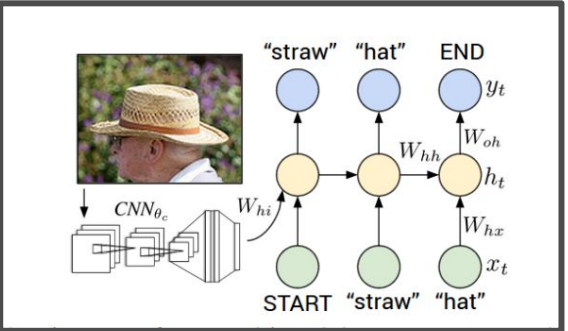
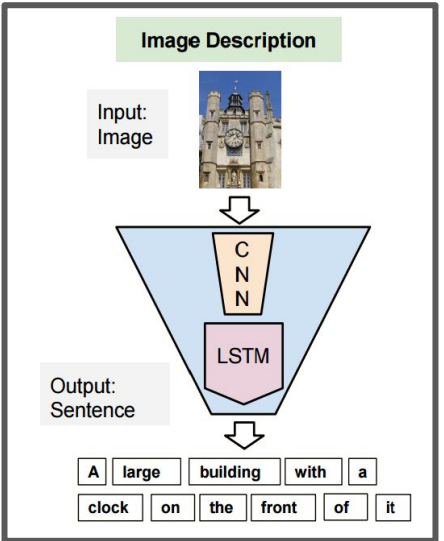
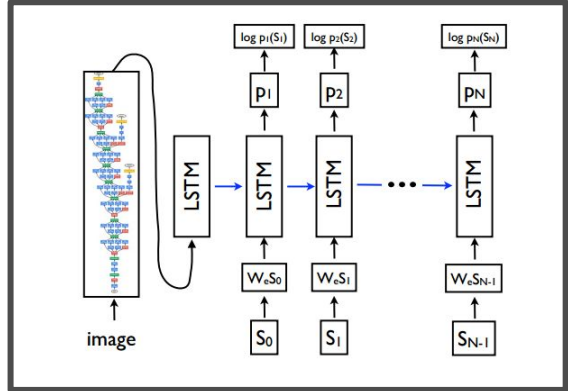
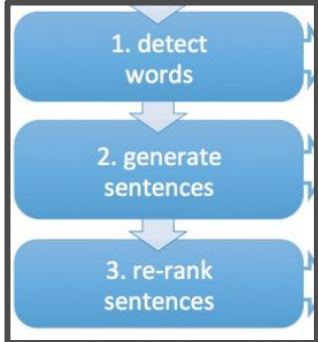


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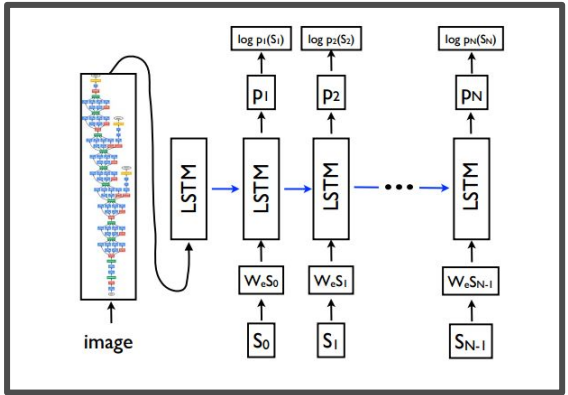


A you
through

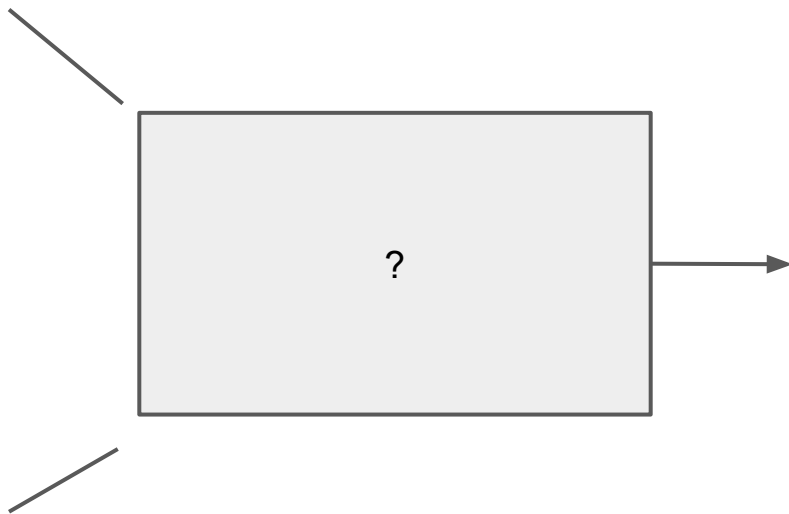
2 The world of neural caption generation models...



[Karpathy and Li 2014]
 [Mao et al. 2014]
 [Vinyals et al. 2014]
 [Kiros et al. 2014]
 [Donahue et al. 2014]
 [Fang et al. 2014]



Neural Image Caption (Vinyal's et. al. 2014)



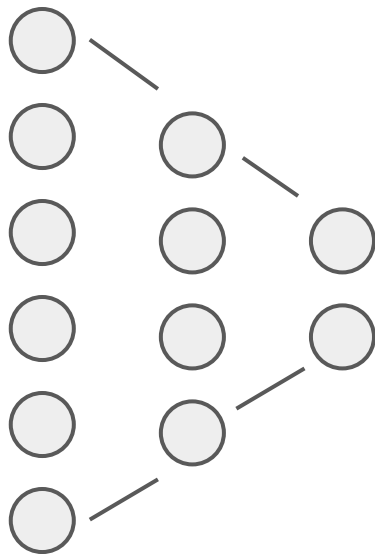
A young girl runs through a grassy field.

Neural Image Caption (Vinyal's et. al. 2014)



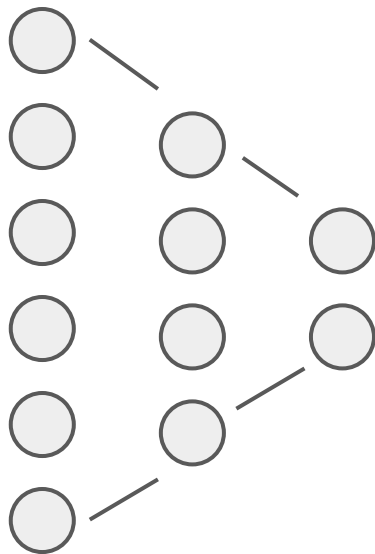
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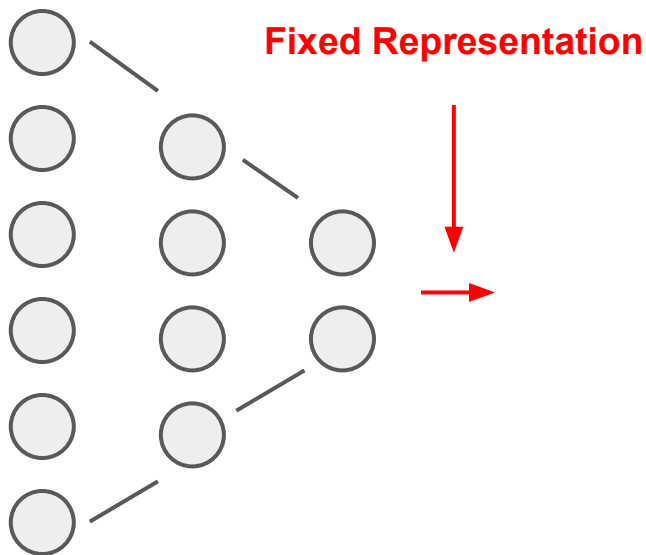
Neural Image Caption (Vinyal's et. al. 2014)



A young girl runs
through a grassy field.

Convolutional Neural Network

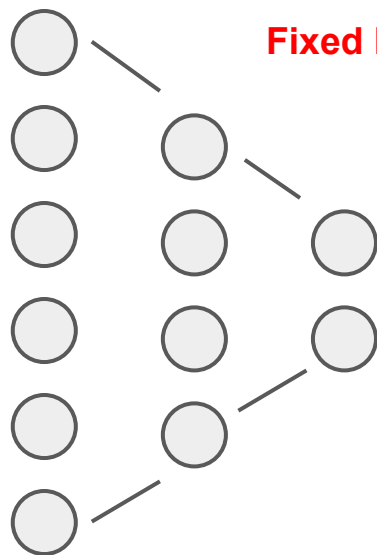
Neural Image Caption (Vinyal's et. al. 2014)



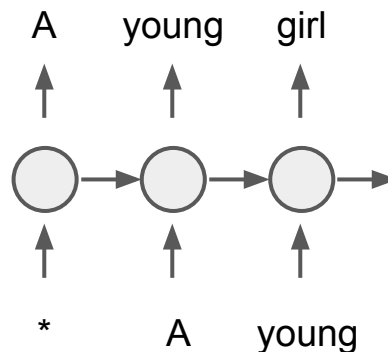
A young girl runs through a grassy field.

Fixed
+
Pretrained Convolutional Neural Network

Neural Image Caption (Vinyal's et. al. 2014)



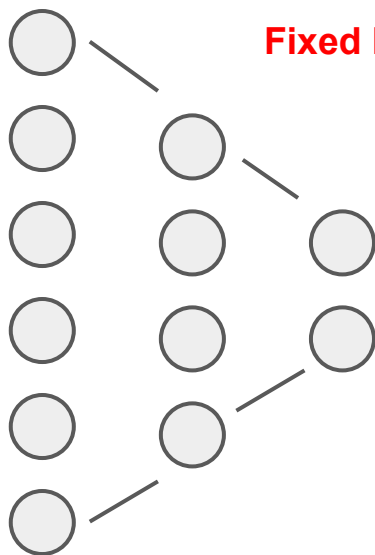
Fixed Representation



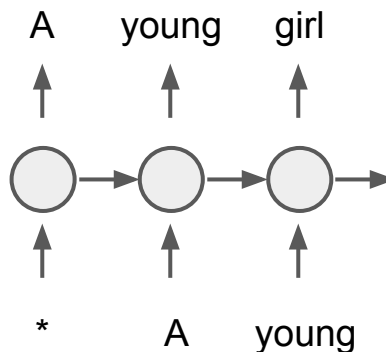
A young girl runs through a grassy field.

Fixed
+
Pretrained Convolutional Neural Network

Neural Image Caption (Vinyal's et. al. 2014)



Fixed Representation



A young girl runs through a grassy field.


Fixed
+
Pretrained
Convolutional Neural Network

Long Short Term Memory
Recurrent Neural Network


Sutskever et al. 2011

The meaning of life is the tradition of the ancient human reproduction: it is less favorable to the good boy for when to remove her bigger. In the show's agreement unanimously resurfaced. The wild pastured with consistent street forests were incorporated by the 15th century BE...

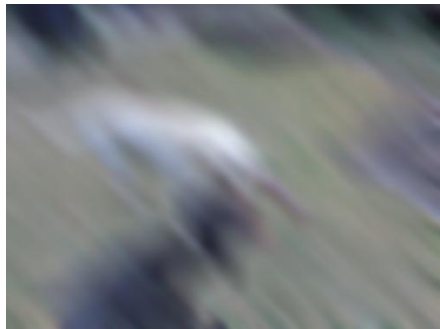
Sutskever et al. 2011

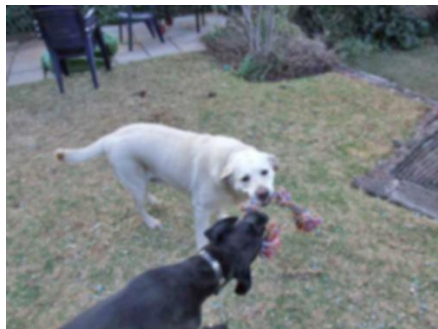
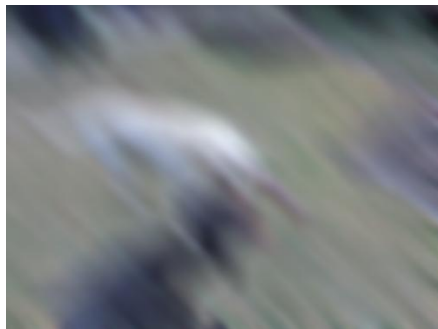
 The meaning of life is the tradition of the ancient human reproduction: it is less favorable to the good boy for when to remove her bigger. In the show's agreement unanimously resurfaced. The wild pastured with consistent street forests were incorporated by the 15th century BE...

Sutskever et al. 2011

 The meaning of life is the tradition of the ancient human reproduction: it is less favorable to the good boy for when to remove her bigger. In the show's agreement unanimously resurfaced. The wild pastured with consistent street forests were incorporated by the 15th century BE...

6

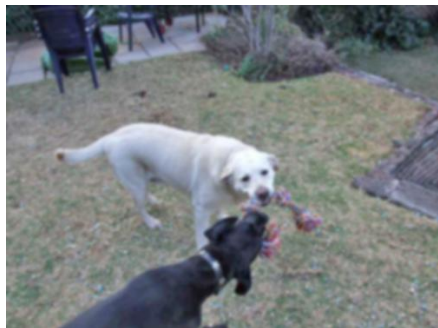
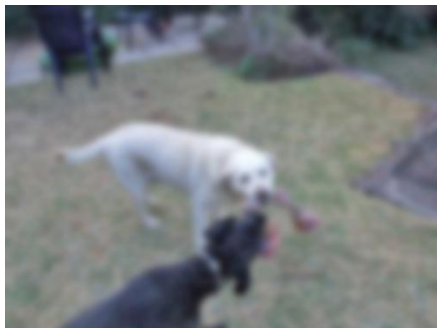
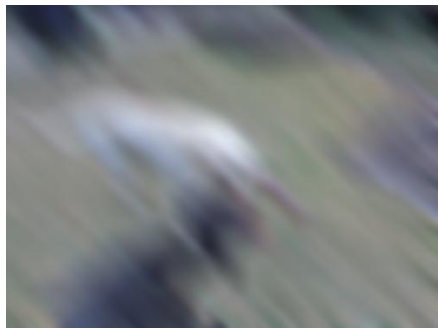




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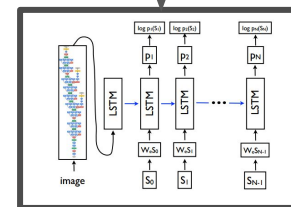
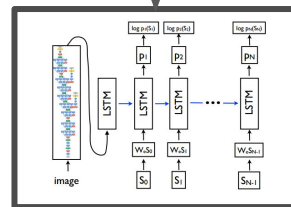
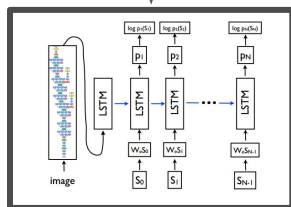
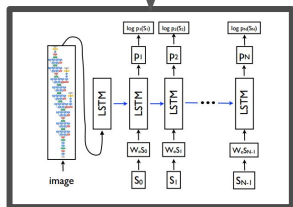
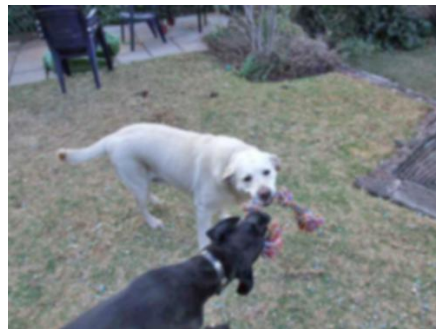
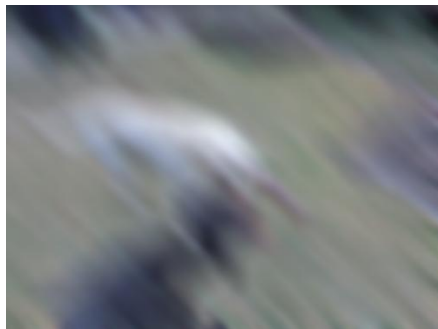


Higher Quality Representations



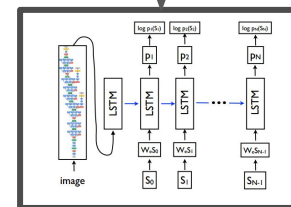
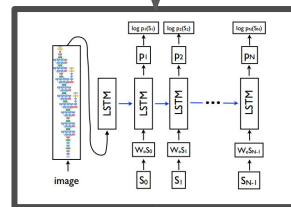
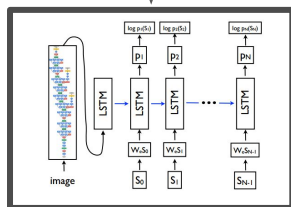
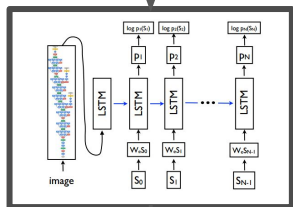
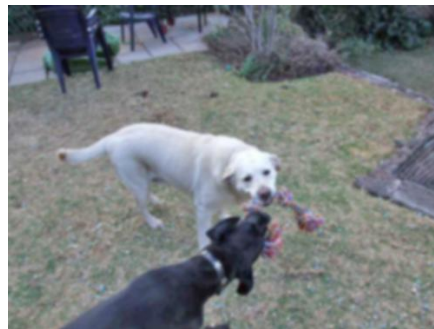
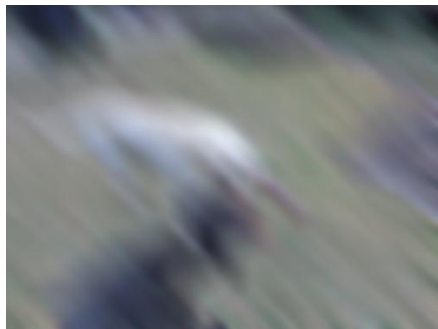


Higher Quality Representations



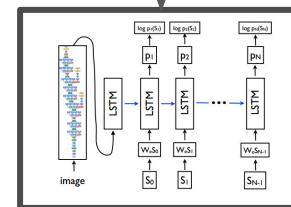
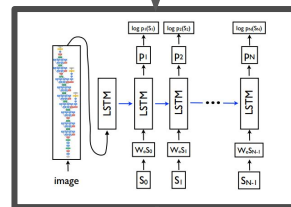
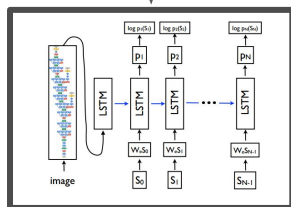
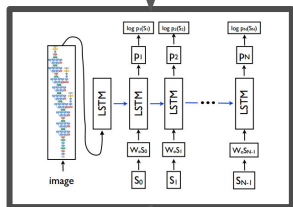
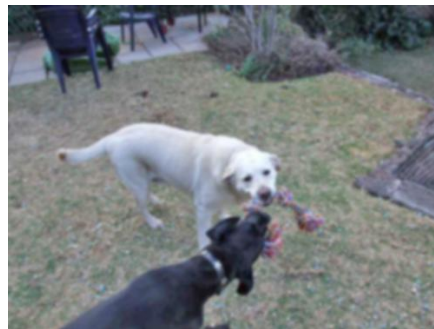
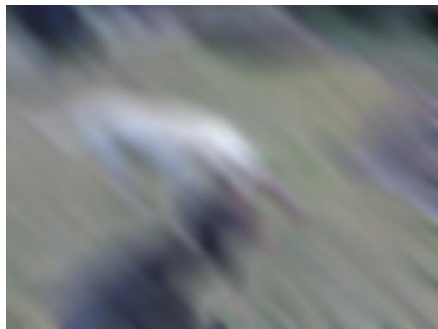


Higher Quality Representations





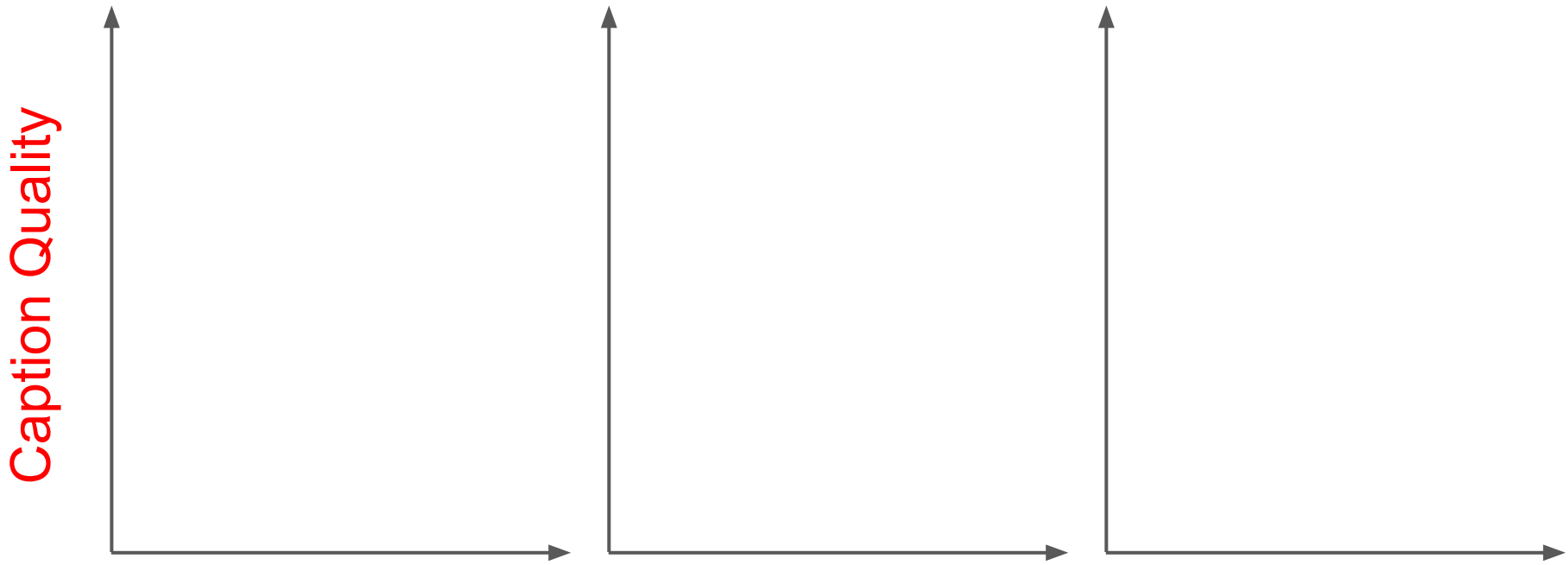
Higher Quality Representations



?? Higher Quality Captions ??

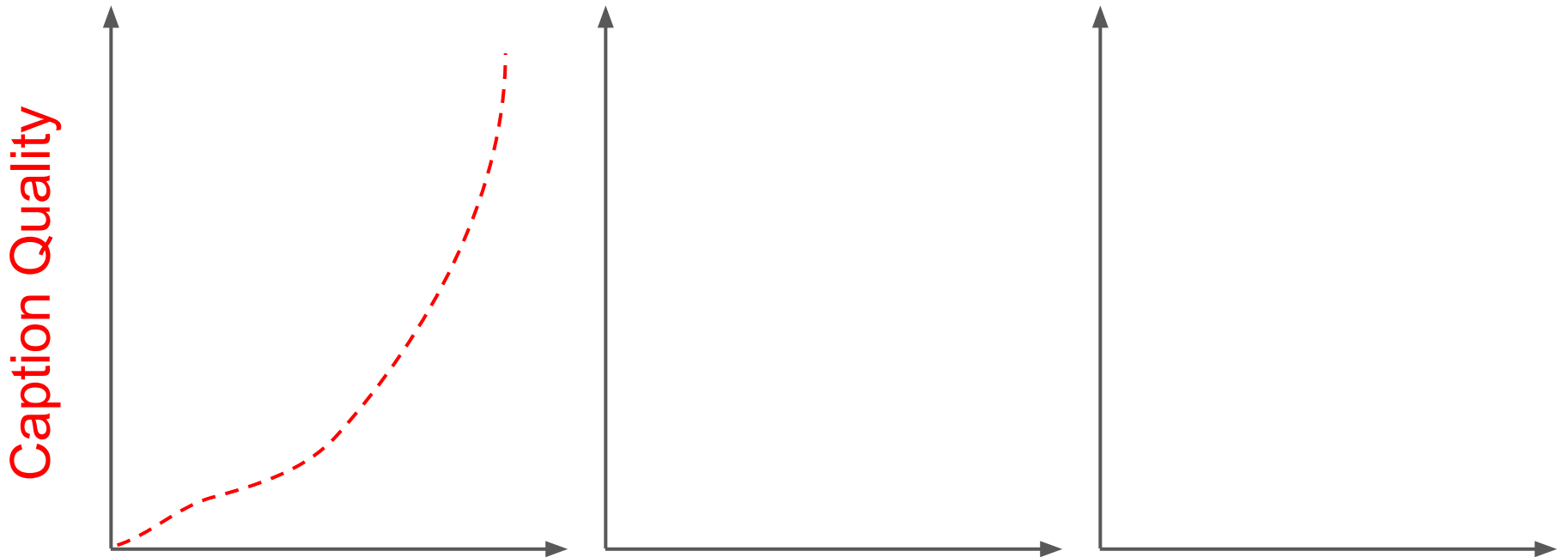


Plausible outcomes....



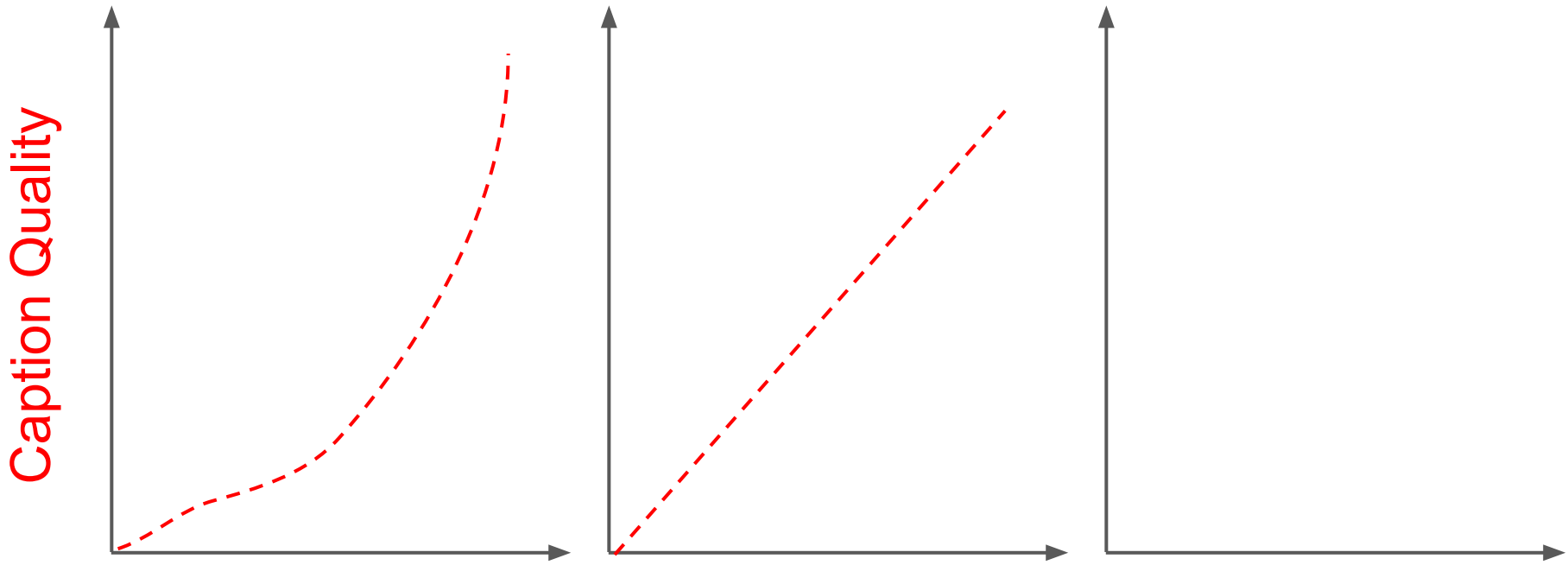
Higher Quality Representations

Plausible outcomes....



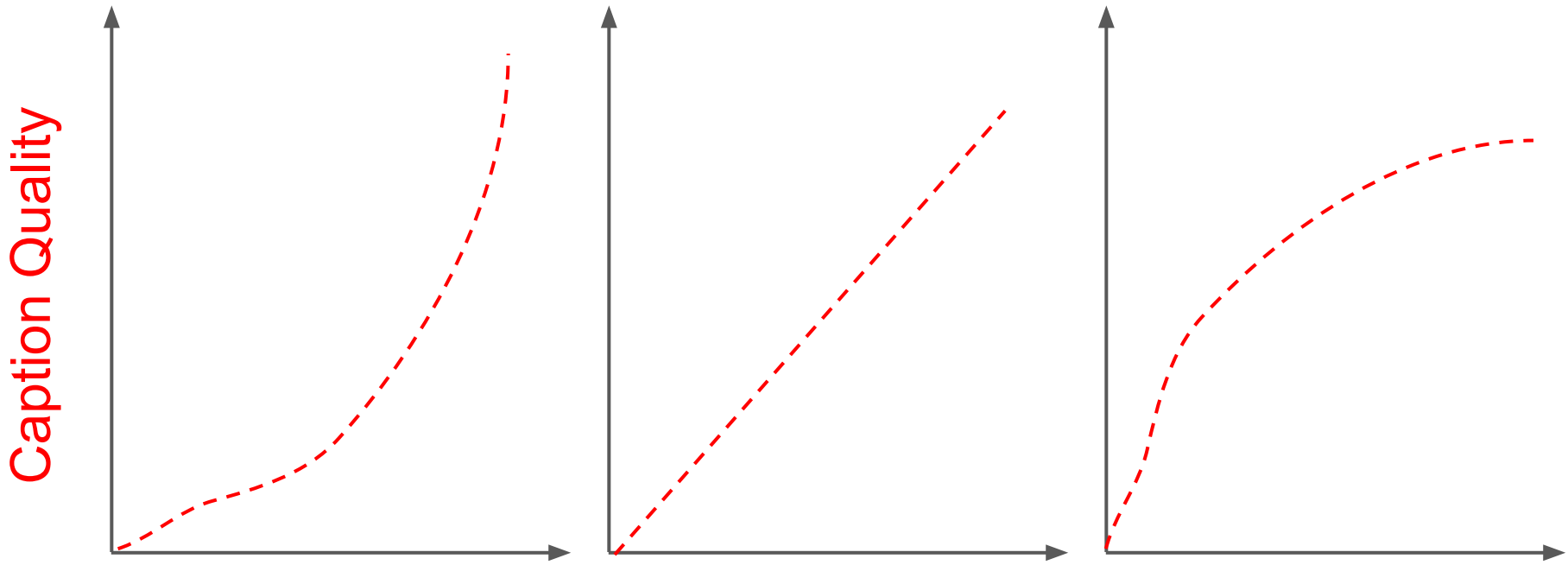
→ Higher Quality Representations →

Plausible outcomes....

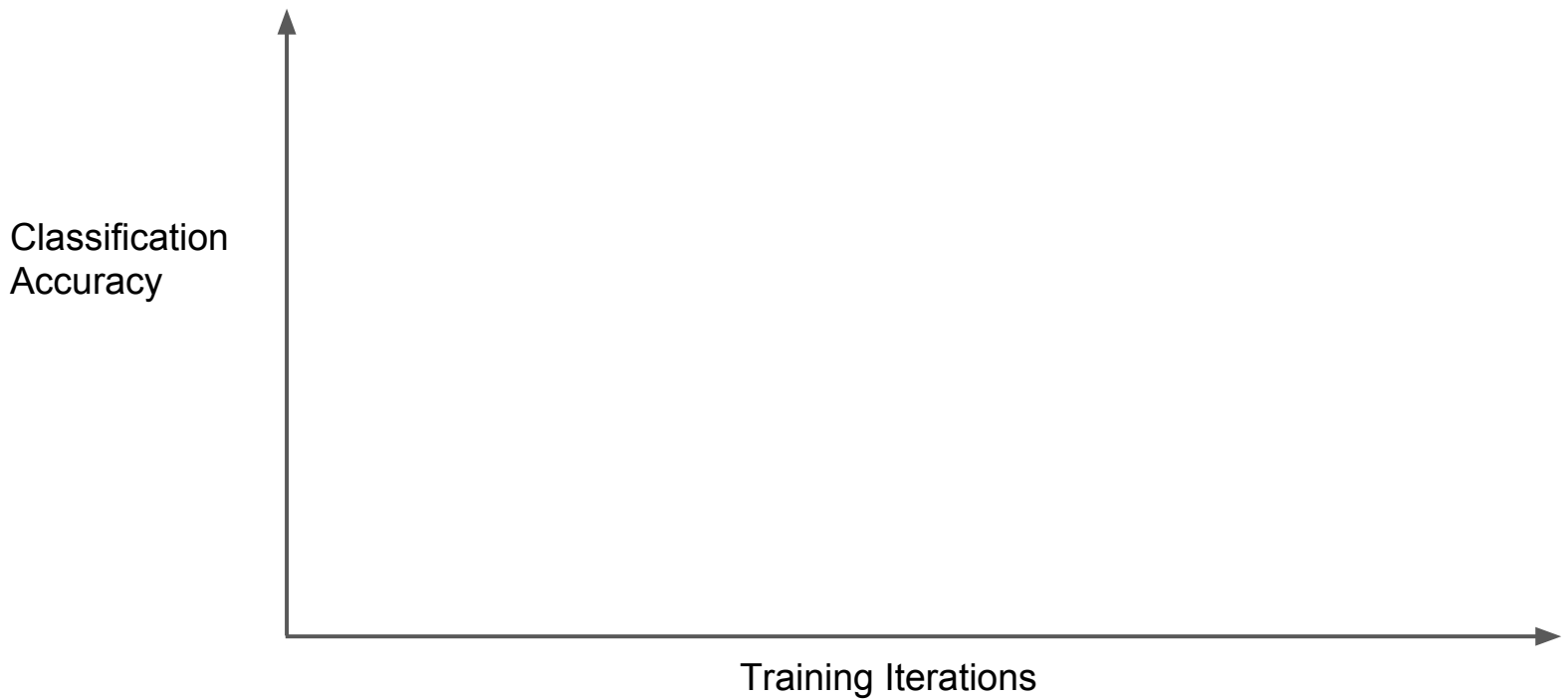


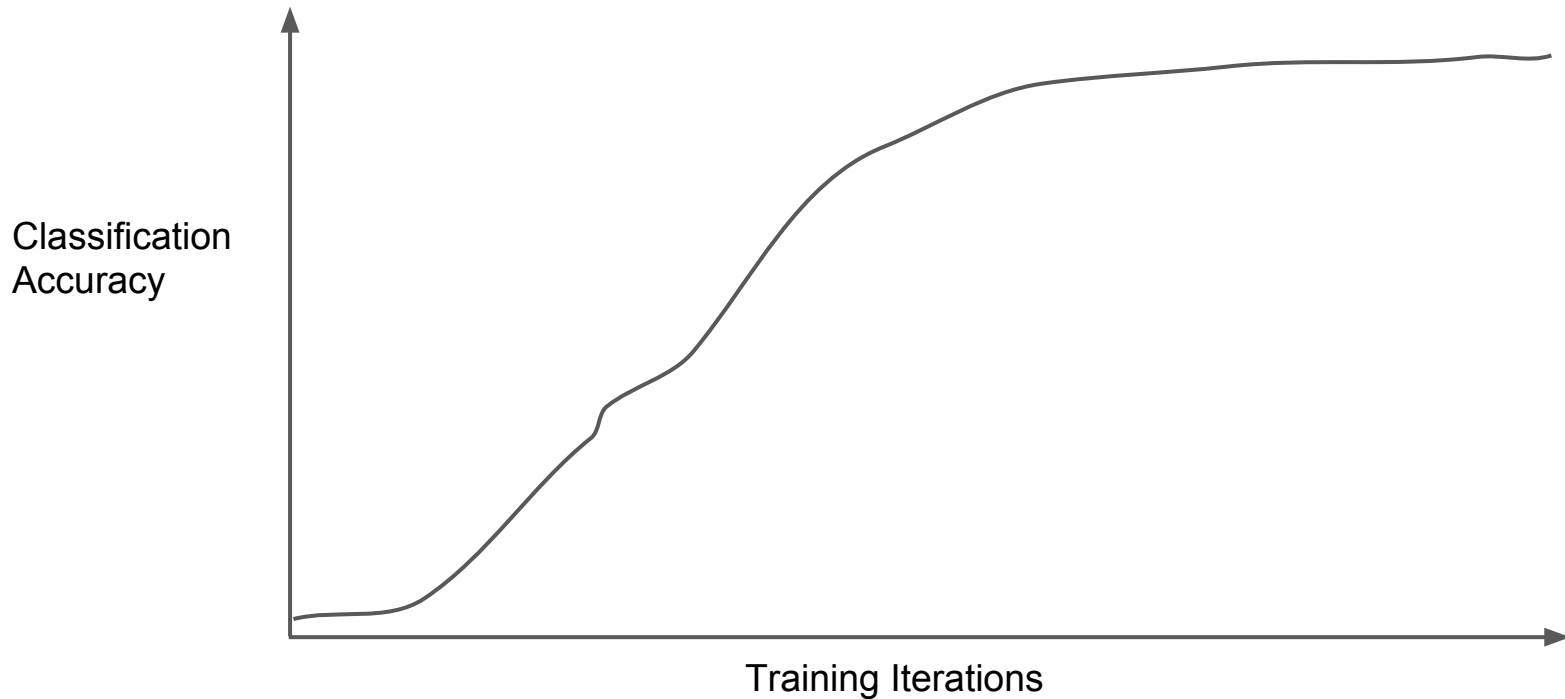
—————▶ Higher Quality Representations —————▶

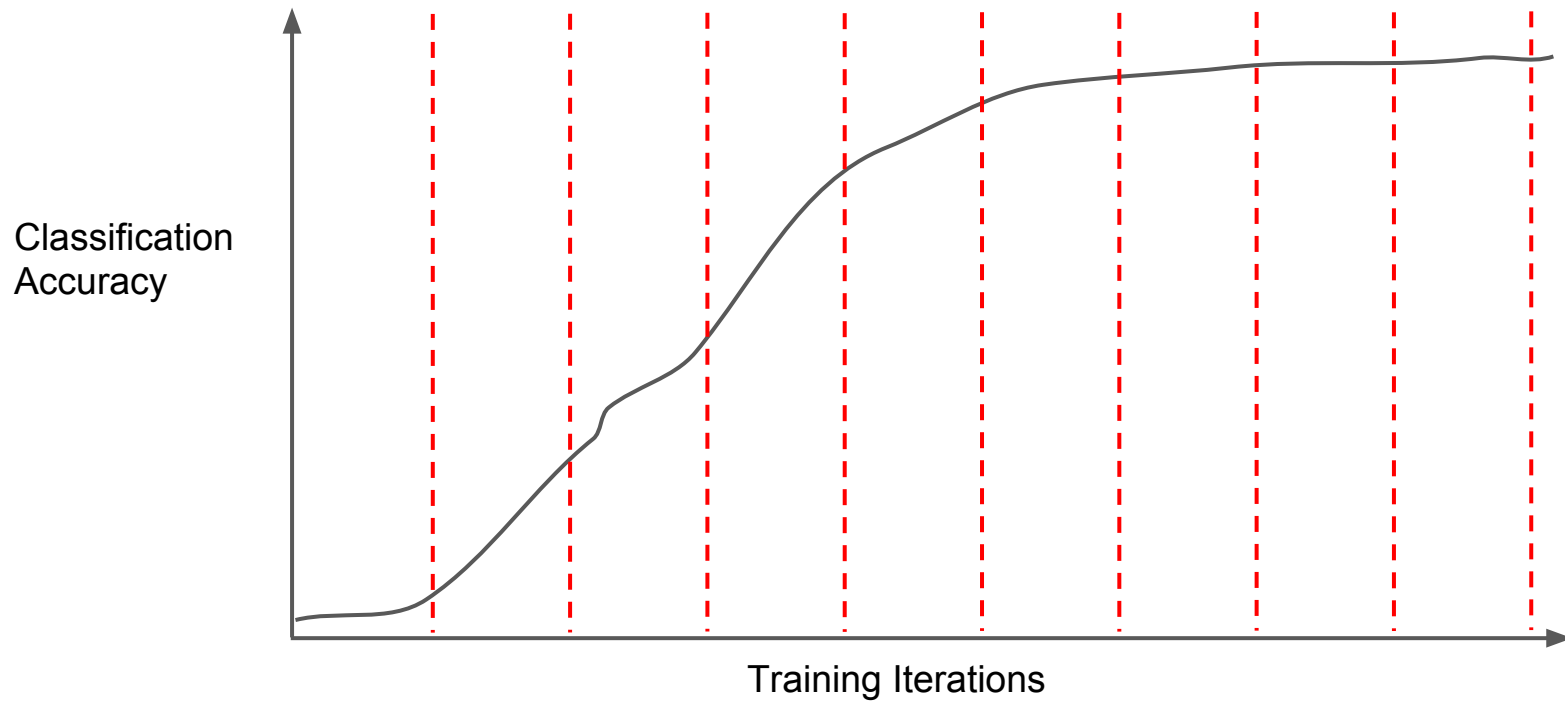
Plausible outcomes....



—————▶ Higher Quality Representations —————▶



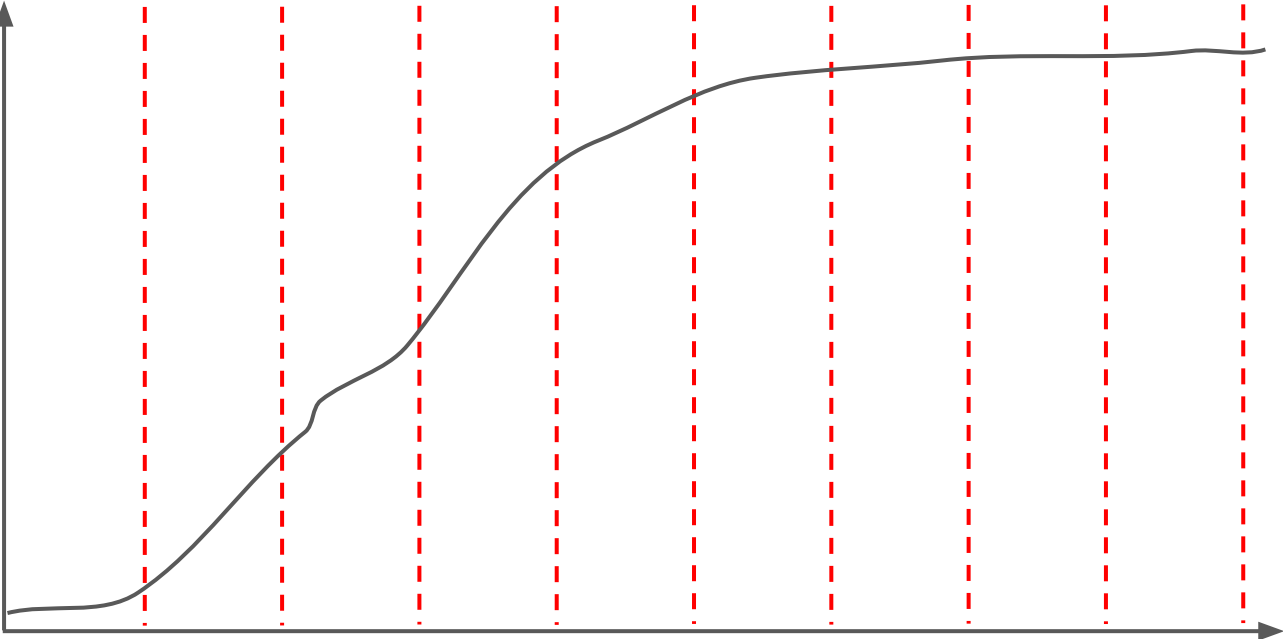
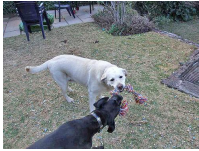
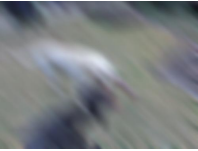




Less "Blur"

Classification Accuracy

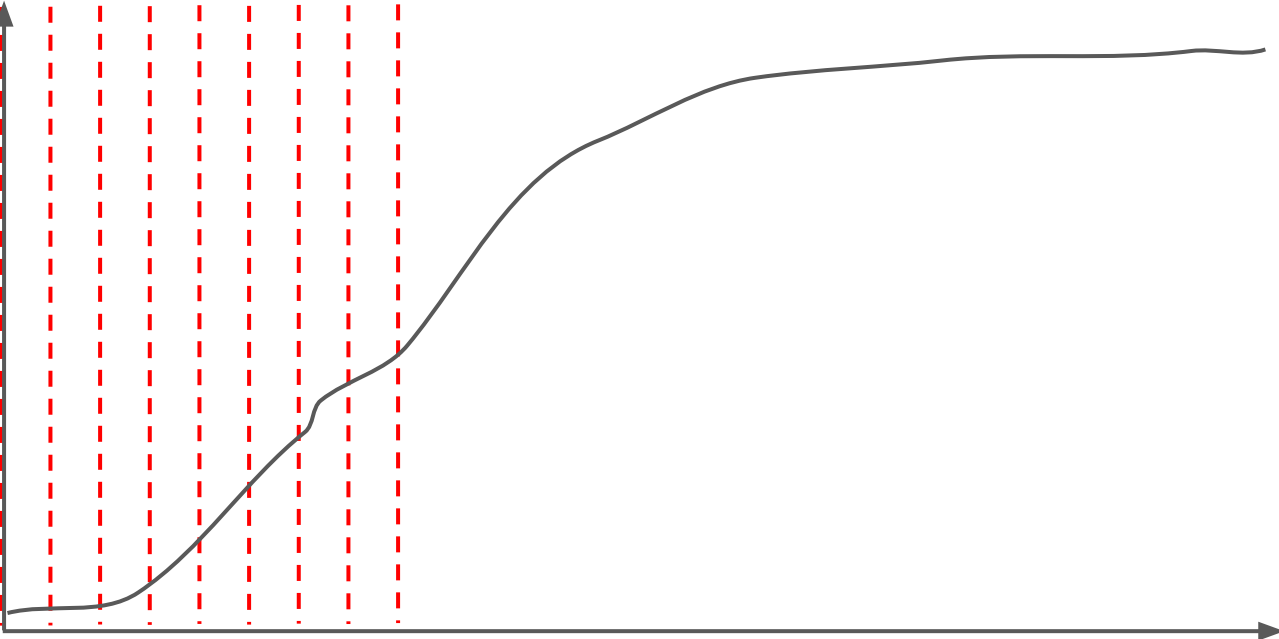
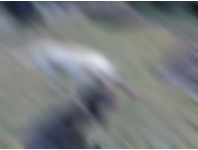
Training Iterations



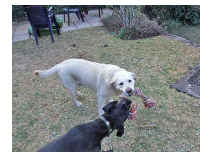
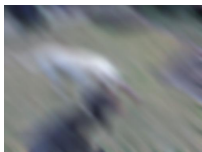
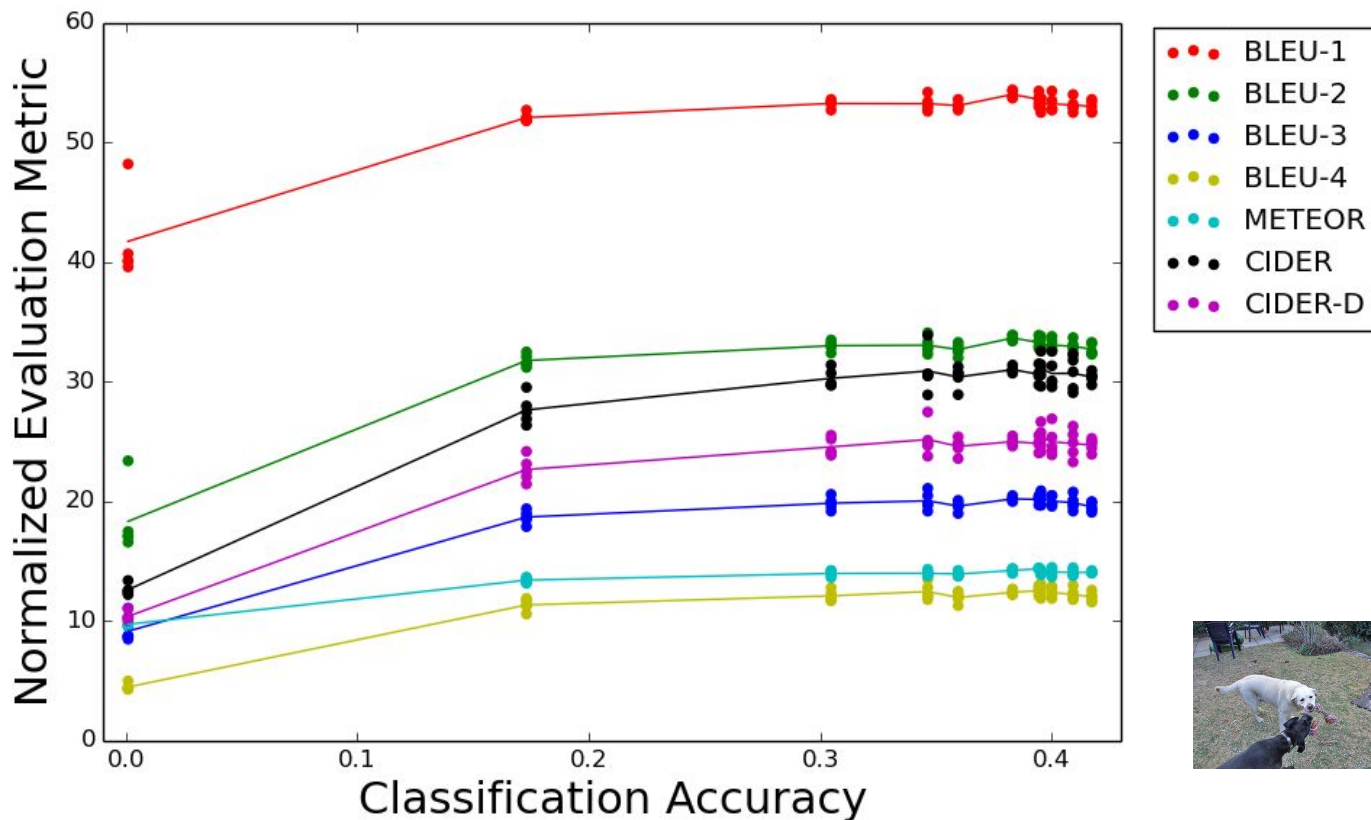
Less "Blur"

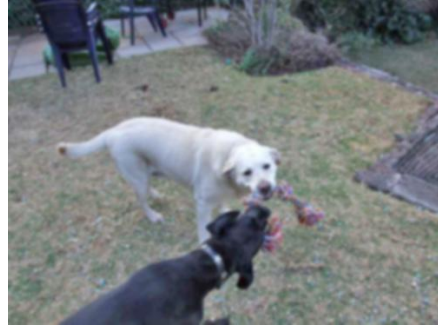
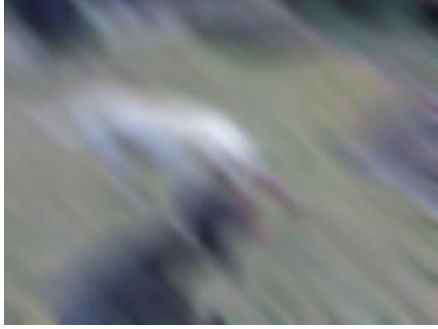
Classification Accuracy

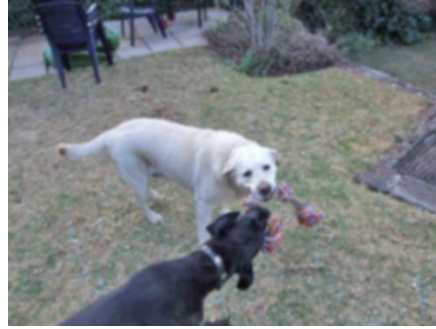
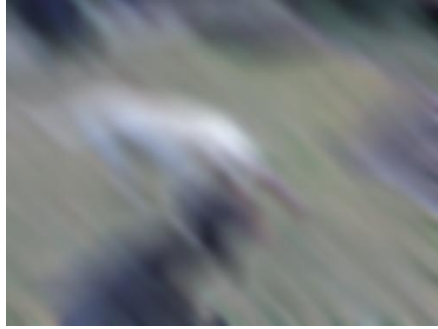
Training Iterations

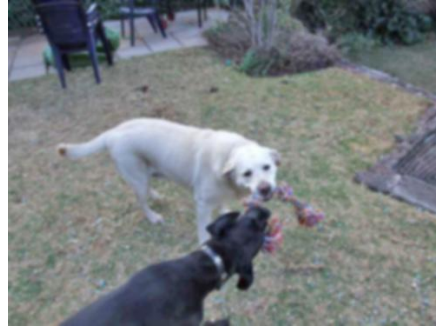
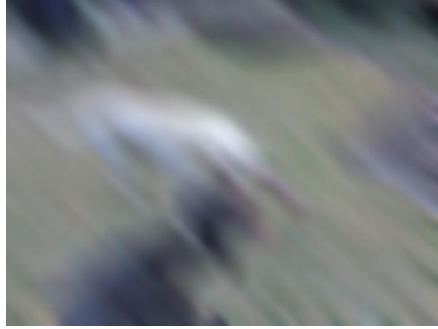


Directly learned: Flickr8k [Hodosh et al. 2013]











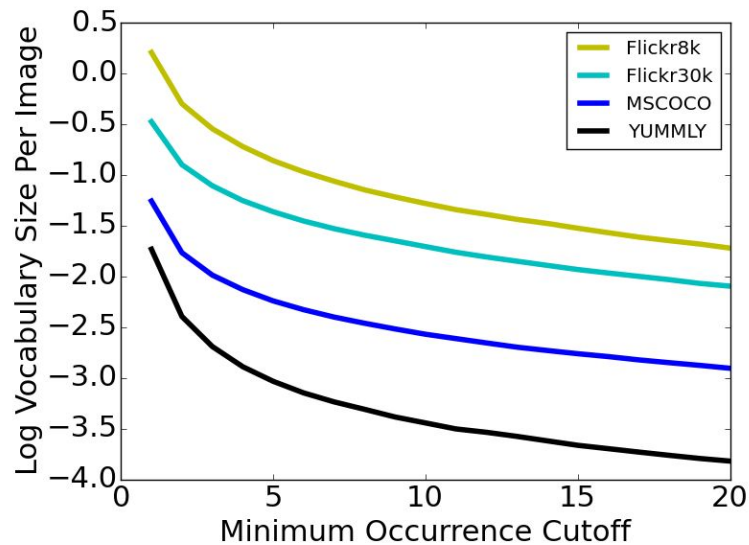


“Malaysian Spicy
Noodles”

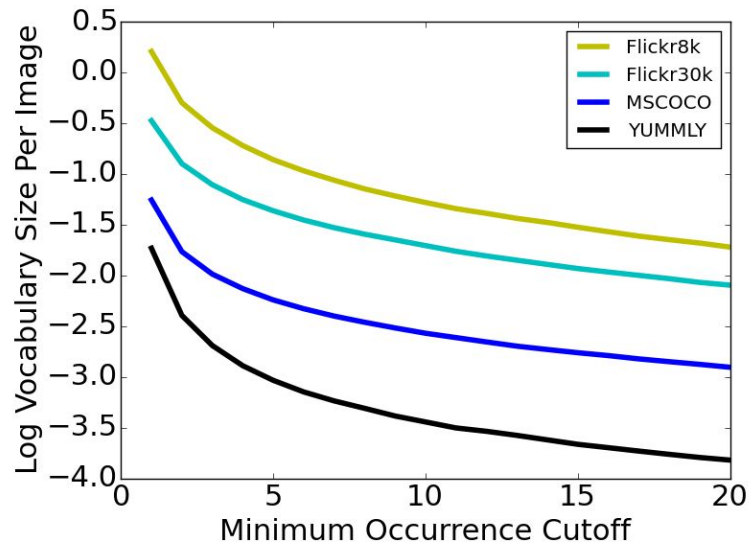
66K Image/recipe pairs, courtesy Yummly.com

Is this really a new *language* task?

Is this really a new *language* task?

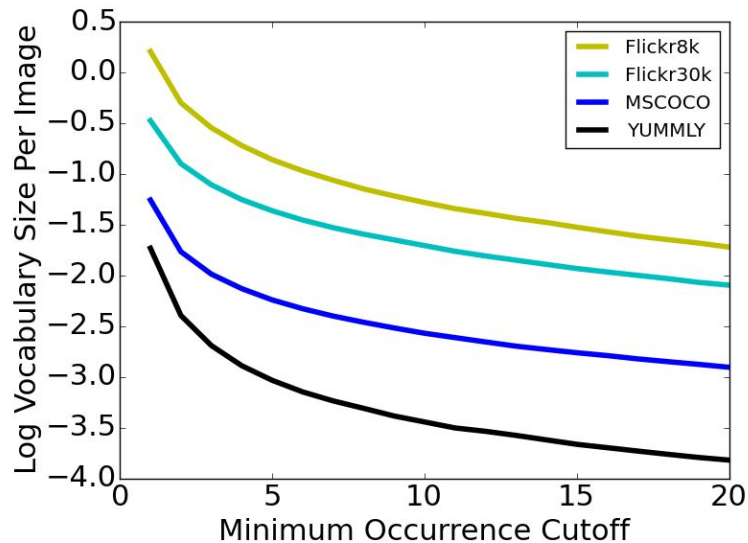


Is this really a new *language* task?



Less variety overall

Is this really a new *language* task?

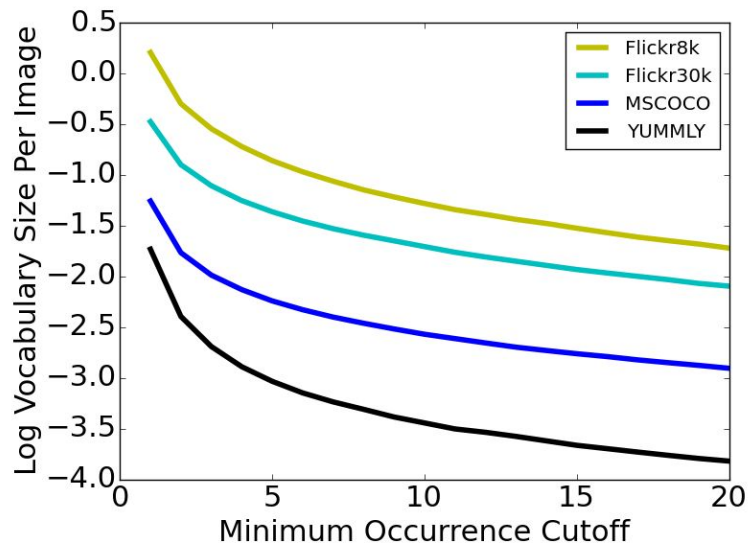


malaysian spicy noodles
chicken curry with lime
cheesy pizza bagels with mushrooms
...



Less variety overall

Is this really a new *language* task?

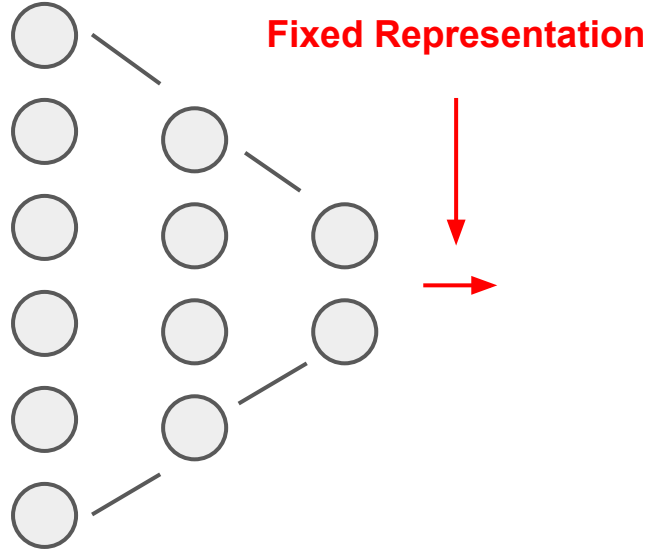


malaysian spicy noodles
 chicken curry with lime
 cheesy pizza bagels with mushrooms
 ...

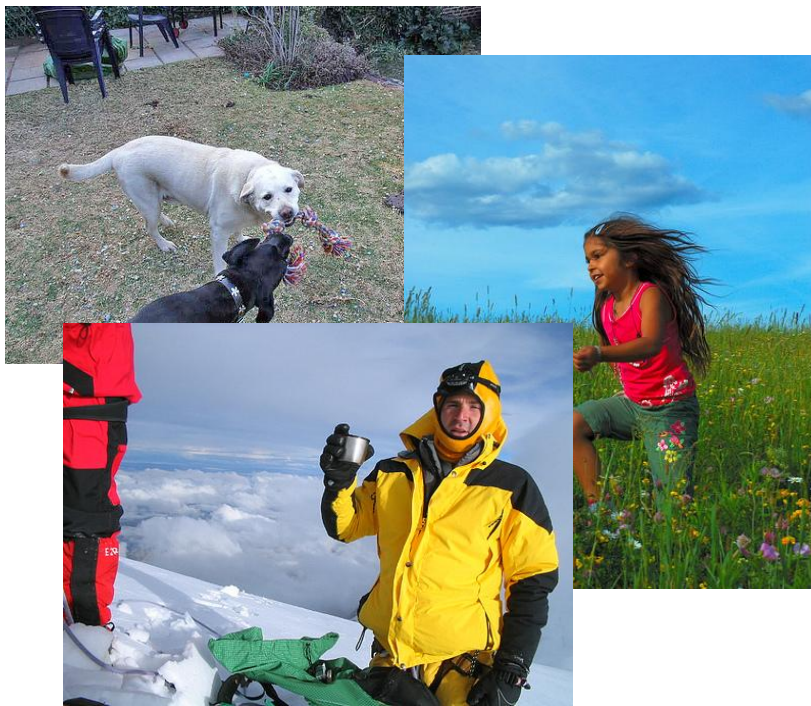


Less variety overall

Shorter captions



**Fixed
+
Pretrained** Convolutional Neural Network

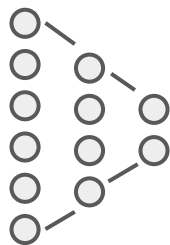


=/=



101K Labeled Food Images in 101 Classes

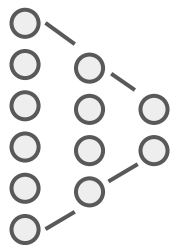
101K Labeled Food Images in 101 Classes



[Krizhevsky et al. 2012]

56.4% Rank-1 Accuracy

101K Labeled Food Images in 101 Classes



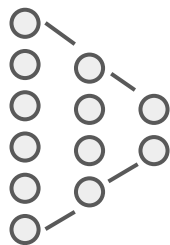
[Krizhevsky et al. 2012]

Transfer Learning



56.4% Rank-1 Accuracy

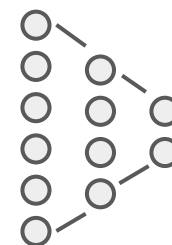
101K Labeled Food Images in 101 Classes



[Krizhevsky et al. 2012]

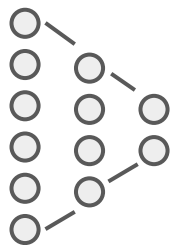
56.4% Rank-1 Accuracy

Transfer Learning



66.8% Rank-1 Accuracy

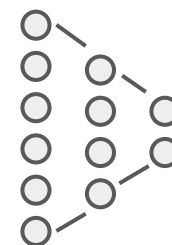
101K Labeled Food Images in 101 Classes



[Krizhevsky et al. 2012]

56.4% Rank-1 Accuracy

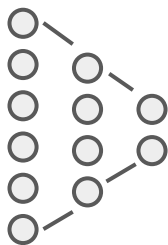
Transfer Learning



[Krizhevsky et al. 2012]

66.8% Rank-1 Accuracy

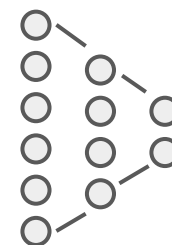
101K Labeled Food Images in 101 Classes



[Krizhevsky et al. 2012]

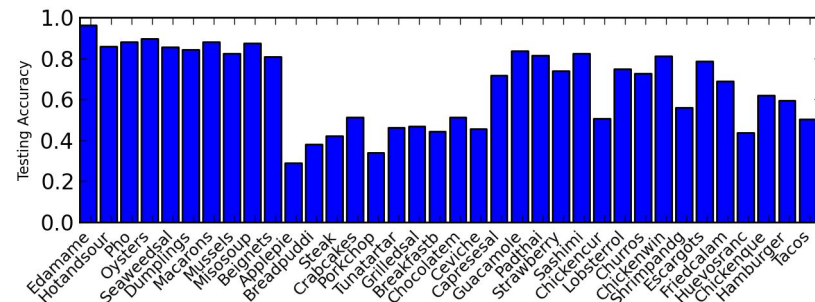
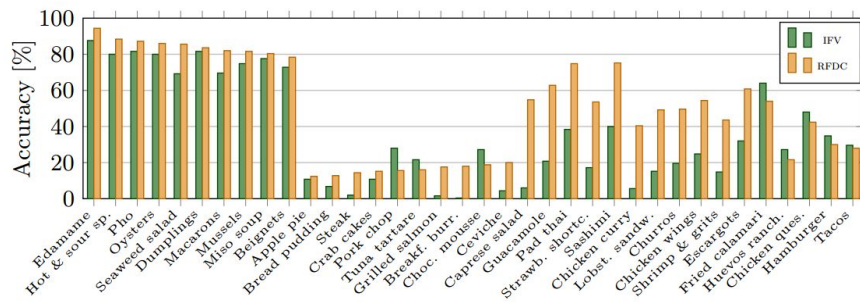
56.4% Rank-1 Accuracy

Transfer Learning

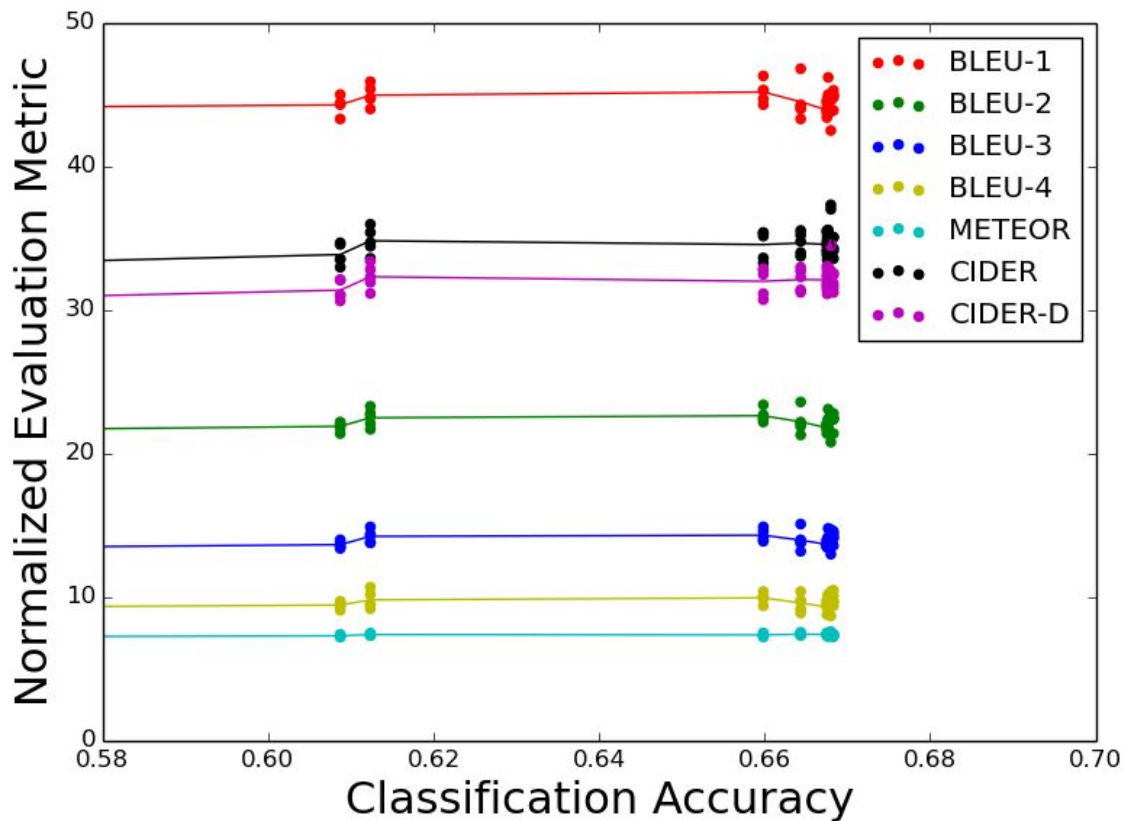


[Krizhevsky et al. 2012]

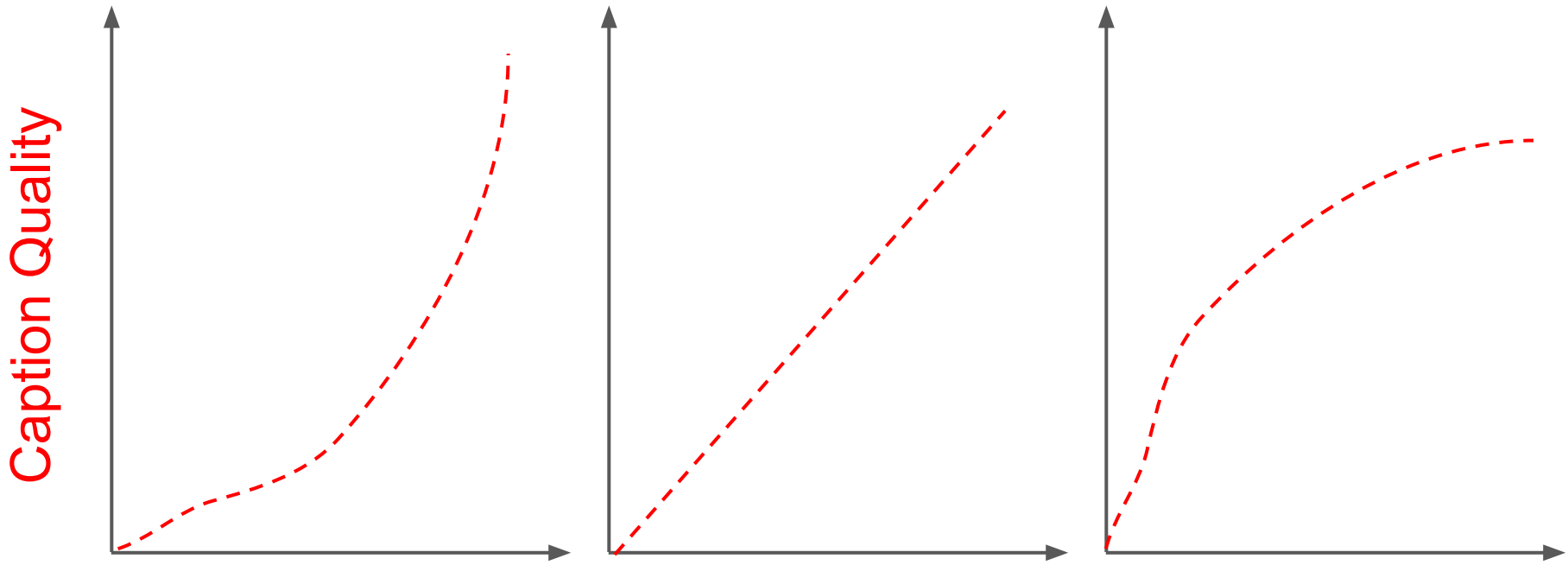
66.8% Rank-1 Accuracy



Transfer learned: Yummly

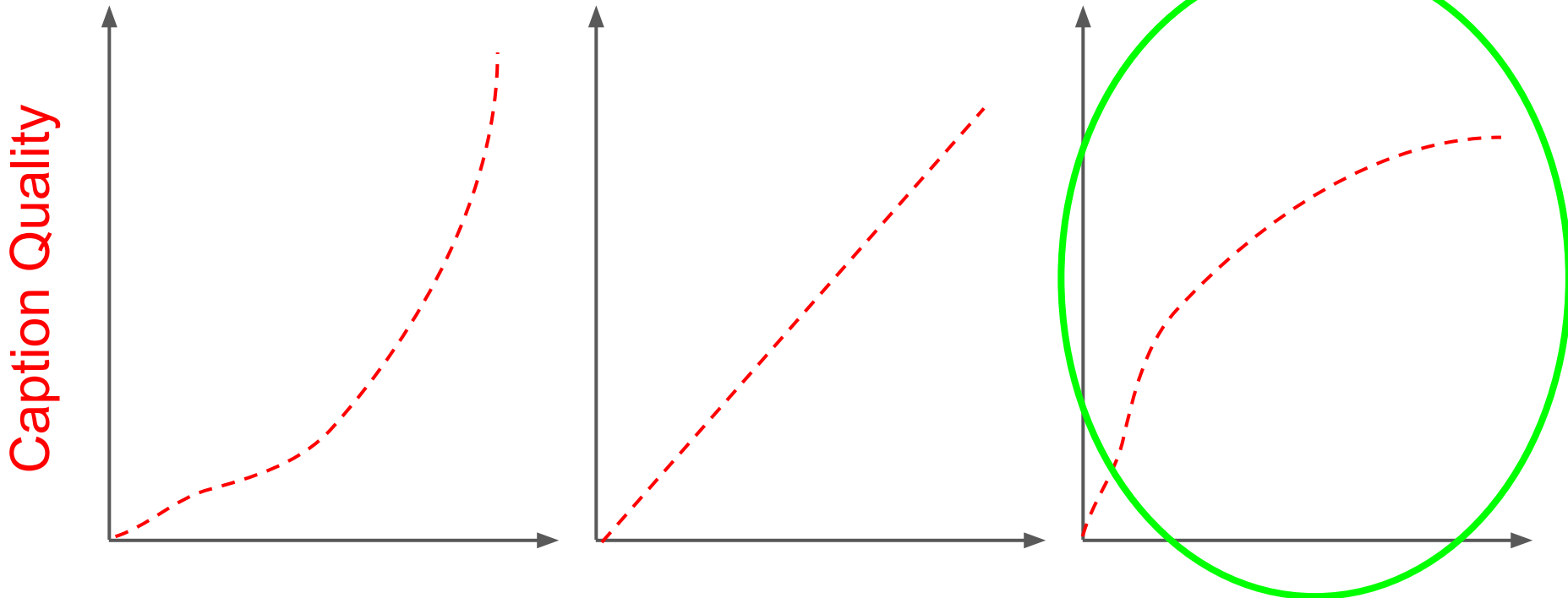


Plausible outcomes....



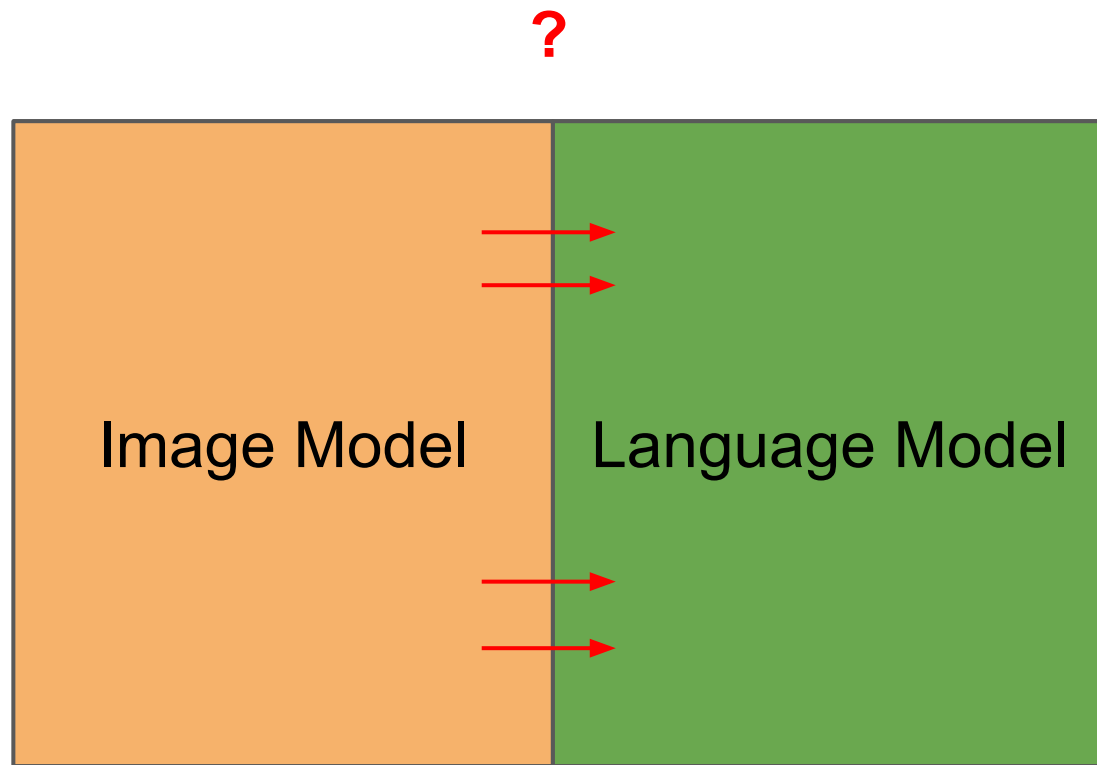
→ Higher Quality Representations →

Plausible outcomes....



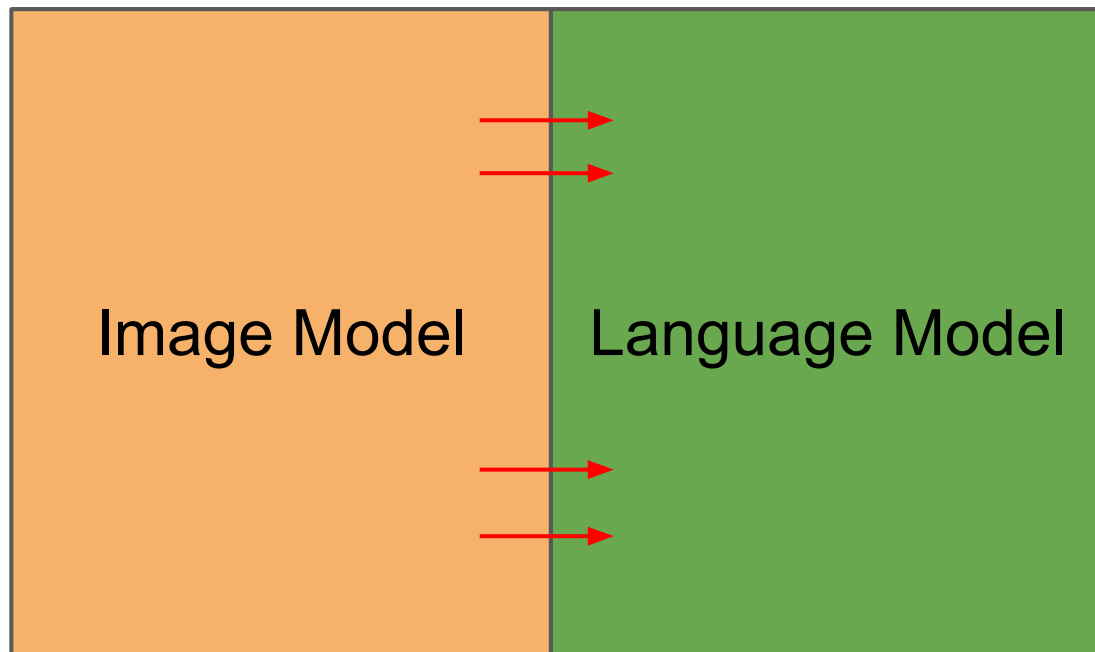
→ Higher Quality Representations →

Implications



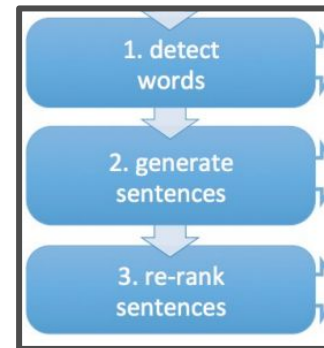
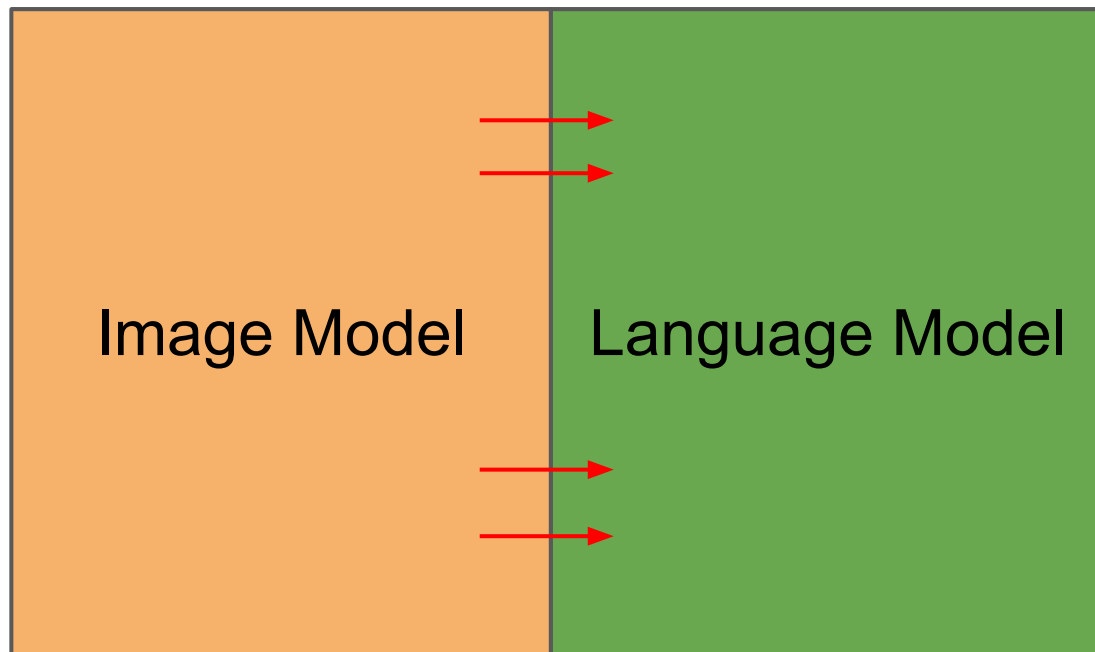
Implications

Mostly Coarse-Grained Information



Implications

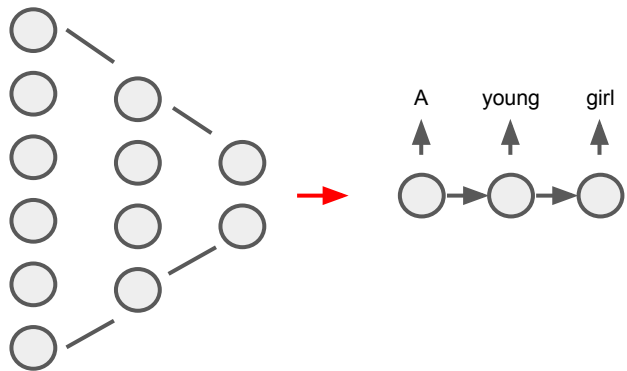
Mostly Coarse-Grained Information



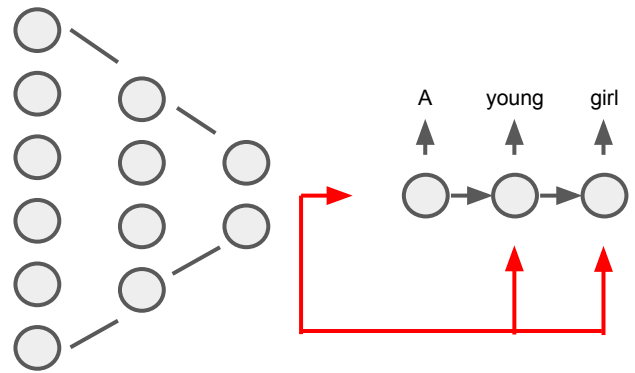
[Fang et al. 2014]

Towards Fine-Grained Captioning

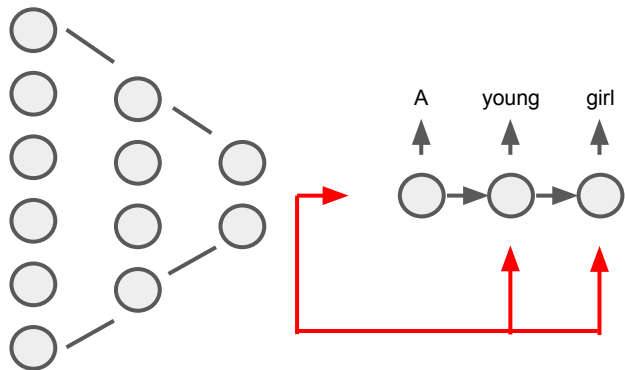
Towards Fine-Grained Captioning



Towards Fine-Grained Captioning

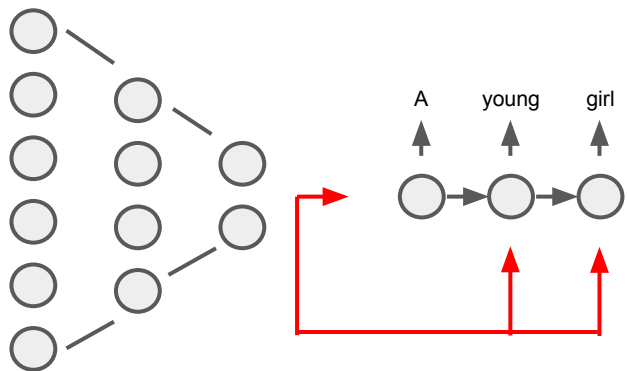


Towards Fine-Grained Captioning



[Lin et al. 2014]

Towards Fine-Grained Captioning



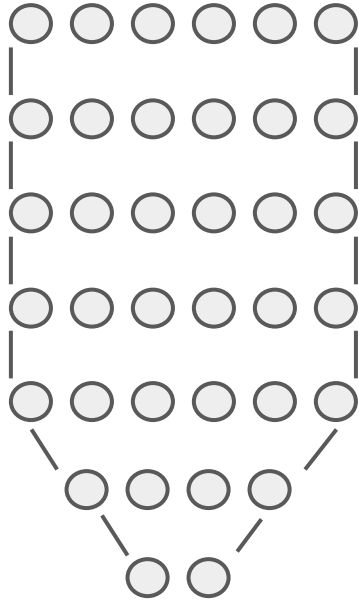
[Lin et al. 2014]

Better Loss Functions!

A few more experiments in the paper...

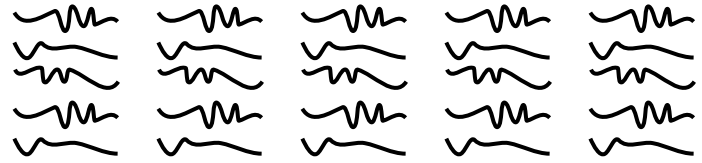
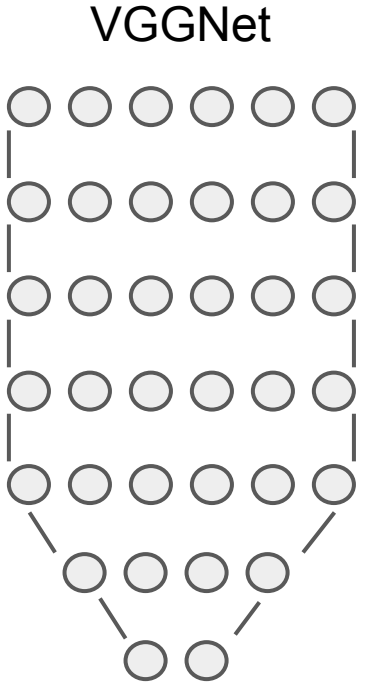
A few more experiments in the paper...

VGGNet



[Simonyan and Zisserman 2014]

A few more experiments in the paper...



vs.



[Simonyan and Zisserman 2014]

Acknowledgements

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Gregory Druck
Abby Lewis
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Serge



Xanda



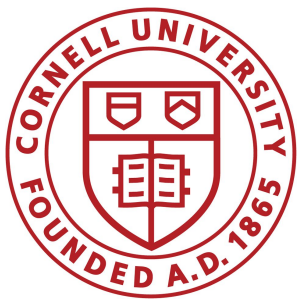
David



Jason



Gregory



CORNELL
NYCTECH



deviled eggs
logprob: -1.52



lasagna
logprob: -4.02



spaghetti carbonara
logprob: -3.81



pulled pork sandwiches
logprob: -3.43



french dip sandwiches
logprob: -5.78



egg salad sandwiches
logprob: -6.05



fish tacos
logprob: -0.27



french toast
logprob: -4.55



tiramisu
logprob: -5.16

Image Representations and New Domains in Neural Image Captioning

Jack Hessel, Nic Saava, Mike Wilber

New domain advice: one caption per image?

	More Captions	More Images
B-1**	55.167	54.243
B-2*	33.567	32.814
B-3	20.633	20.300
B-4	13.133	13.014
METEOR	13.105	13.096
CIDEr	21.428	20.418
CIDEr-D	16.350	15.550
Proportion Unique**	14.8%	9.96%
Training Perplexity**	14.69	16.01
Validation Perplexity*	25.86	25.33

AlexNet vs VGG-Net

	AlexNet	VGG
Top-1 ImageNet Val Acc	57.1%	75.6%
B-1	54.187	53.913
B-2	33.967	33.527
B-3**	20.640	20.007
B-4**	12.833	12.213
METEOR	14.559	14.559
CIDEr	32.416	31.362
CIDEr-D*	26.200	25.242
Proportion Unique***	20.5%	17.0%
Training Perplexity***	10.79	11.04
Validation Perplexity***	17.84	17.66