

OVERVIEW

- **Problem:** in a GUI, click on object in image -> get its outline (useful for image editing).



- 'DeepMask' is great at outputting the outlines of all objects within an image from its pixels[1].
- 'DeepCrop' is a modification of DeepMask which adds a 'pixel click' input via a 4th image plane that encodes distance from the click coordinate, to **crop clicked object** from image.
- Two versions of DeepCrop are presented, with one slightly outperforming DeepMask.

BACKGROUND

- DeepMask generates an object mask (outline) for images with a single centered object.
- The image is put through a feature extraction network, and then a 'mask' branch generates the outline and a 'score' branch scores the likelihood the input contains a centered object.
- For segmenting multiple non-centered objects in any image, the model is **applied as a convolution and the highest scoring masks are kept**.

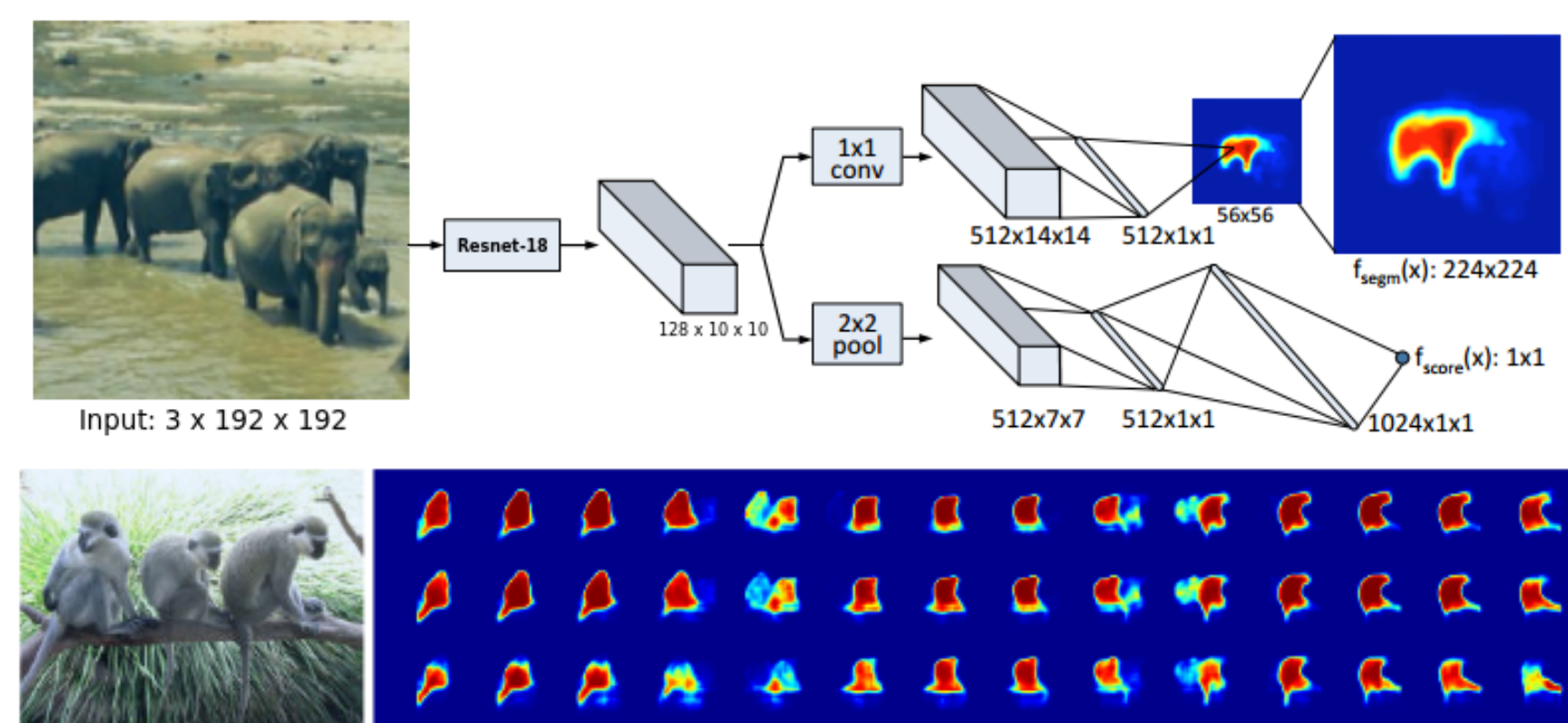


Figure 1: Diagram of the DeepMask model and an example convolution across an image

- The MS COCO dataset is used [3]. It contains a total of 123,287 images with 886,284 objects, but only 80 categories with outlines.



Figure 2: Examples from COCO

DEEPCROPV1

- V1 approach is to revise DeepMask by **removing the score branch and adding three measures of distance from the 'click' pixel as additional inputs to the mask branch**.
- The intuition is that providing a click coordinate removes need to assume object centrality and allows directly processing the entire image.
- This was implemented as a **direct modification to the open source DeepMask code, in Torch**.

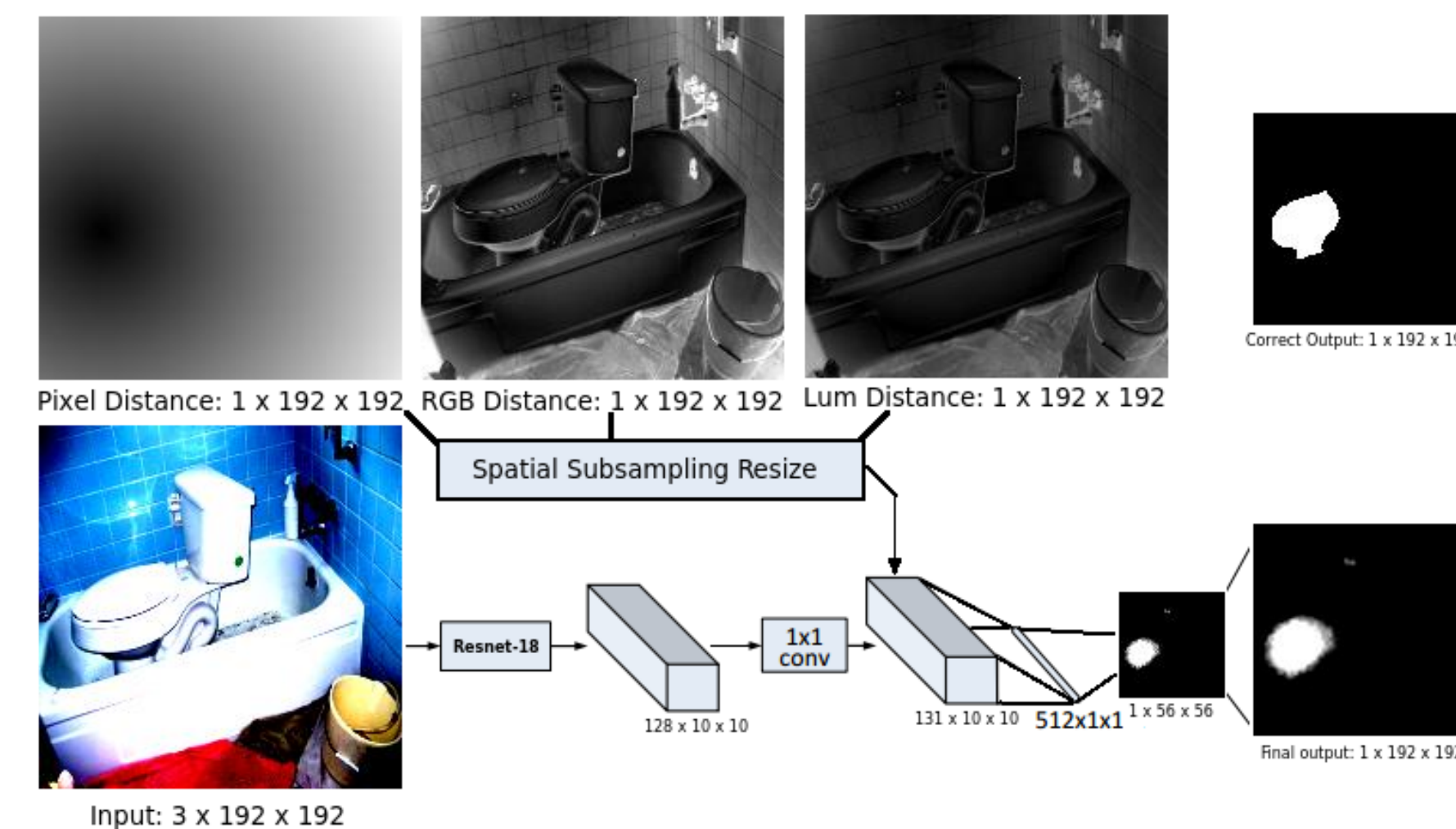


Figure 3: Visualization of the 'DeepCrop1' model

DEEPCROPV2

There are several problems with DeepCropV1:

- The input image is significantly downsized and so there is a limit to the degree of accuracy the output can have.
- Having to segment a non-centered object presents additional complexity
- The feature extraction net is not provided the click location and so cannot use it.

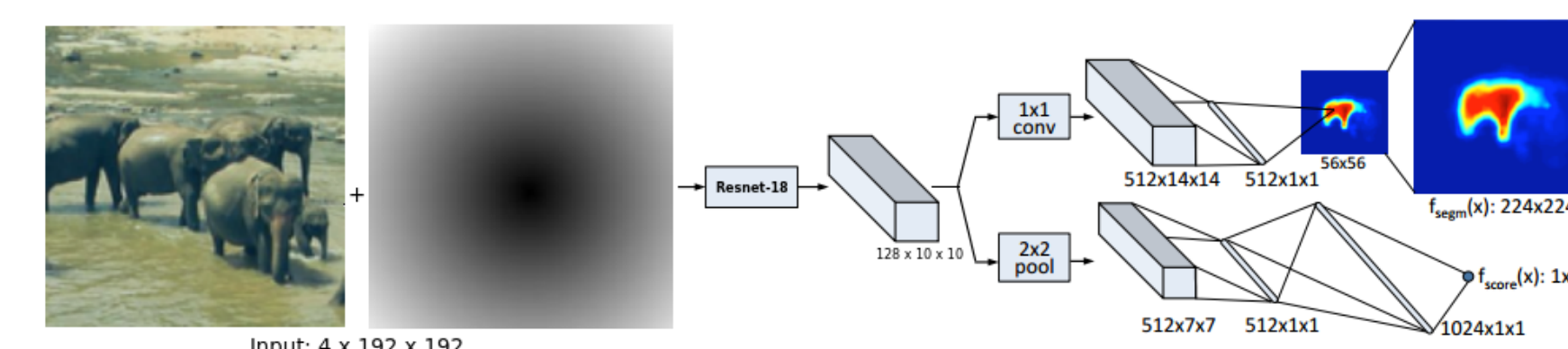


Figure 4: Diagram of the DeepCropV2

- A second version of DeepCrop was developed to take this into account. This revised model is **the same as DeepMask, but the inputs it is trained on include a fourth image plane that represents distance from the clicked pixel**.
- This 'DeepCropV2' works significantly better than DeepCropV1, and also slightly outperforms DeepMask.

RESULTS



Figure 5: Train graphs. An epoch is 4000 batches, and a batch has 16 examples. DeepMaskV2_50 has a larger feature extraction net and starts with weights copied from a trained DeepMask model.

Model	Metric	Train			Test		
		Mean IoU	Median IoU	IoU@0.7	Mean IoU	Median IoU	IoU@0.7
DeepMask		61.3%	69.0%	48.5%	60.4%	68.16%	47.0%
DeepMask-50		68.0%	75.1%	59.9%	65.8%	73.1%	55.4%
DeepCropV1		15.5%	0.0%	8.0%	11.6%	0.0%	4.1%
DeepCropV2		63.8%	70.0%	50.1%	63.5%	69.6%	49.3%
DeepCropV2-50		67.5%	74.6%	58.2%	66.1%	73.2%	55.3%

Table 1: Comparison of models. 'IoU' is Intersection-over-Union, IoU@0.5 is IoU recall with 50% threshold. Inputs are as in training. Both training and test values were evaluated with 5000 samples from their respective sets.

- The metrics show that the **DeepCropV2 model outperforms DeepMask slightly**, and the larger '-50' models perform best.
- A reasonable conclusion is that **there is not a**

significant difference due to both models being trained and tested on centered inputs, and a comparison on outlining objects in whole images would yield a more significant difference.

DISCUSSION

- The project presented several challenges: understanding DeepMask, learning to program in Torch, and the sheer amount of compute power these models require (3 AWS GPU instances were used for a total of \$1000 AWS credit granted by Amazon for this project)
- **Overall I consider the project successful**, since

DeepCropV1 provided interesting results as to the limitations of DeepMask and DeepCropV2 was an actually useful model.

- I expected the outcome, due to the outlined problems with DeepCropV1 and DeepCropV2 being the same as DeepMask but with additional input.

FUTURE

There are three major areas for future work:

- Evaluating DeepCrop for whole images.
- Integrating DeepCrop with the segmentation-refinement model SharpMask[2].
- Further developing a GUI to test the usefulness of DeepCrop.

REFERENCES

- [1] Pedro O. Pinheiro et al. Learning to segment object candidates. In *NIPS*, 2015.
- [2] Pedro O. Pinheiro et al. Learning to refine object segments. In *ECCV*, 2016.
- [3] Tsung-Yi Lin et al. Microsoft COCO: common objects in context. *CoRR*, abs/1405.0312, 2014.