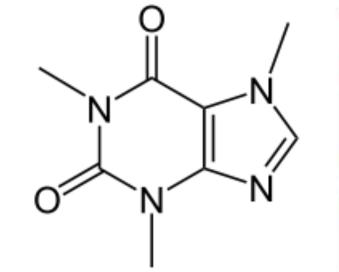
CAFFE: CONVOLUTIONAL ARCHITECTURE FOR FAST FEATURE EMBEDDING

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caffe.berkeleyvision.org

Abstract

Caffe provides vision scientists and practitioners with

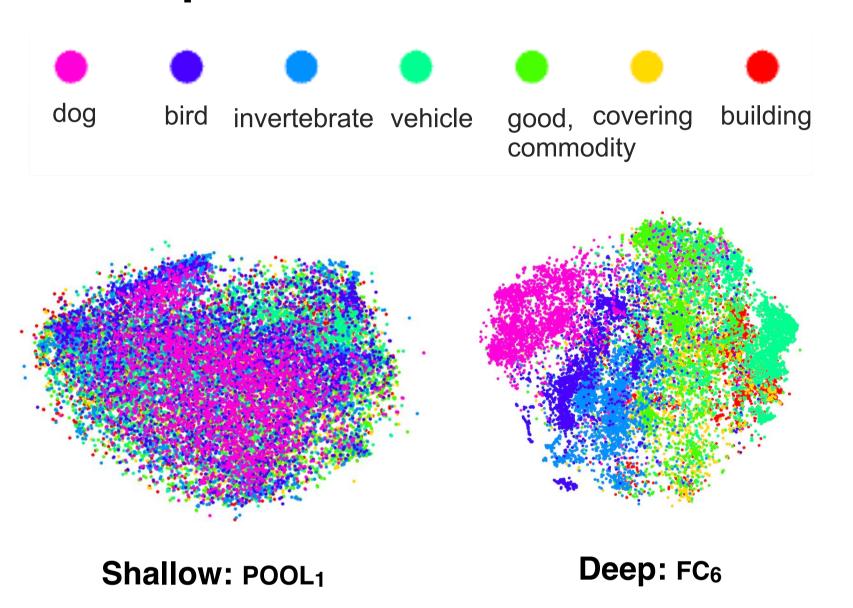
- a framework for state-of-the-art deep learning algorithms
- a collection of reference models
- processing of > 40 million images a day
 ≈ 2.5 ms per image on a single GPU
- Python and MATLAB bindings for experiments

in a fully open-source BSD-licensed C++ / CUDA library.

Framework Comparison

Framework	License	Core language	$\operatorname{Binding}(s)$	CPU	GPU	Open source	Training	Pretrained models	Development
Caffe	BSD	C++	Python, MATLAB	✓	✓	✓	✓	✓	distributed
cuda-convnet [7]	unspecified	C++	Python						discontinued
Decaf [2]	BSD	Python						\checkmark	discontinued
OverFeat [9]	unspecified	Lua	C++,Python					\checkmark	centralized
Theano/Pylearn2 [4]	BSD	Python				/			distributed
Torch7 [1]	BSD	Lua					✓		distributed

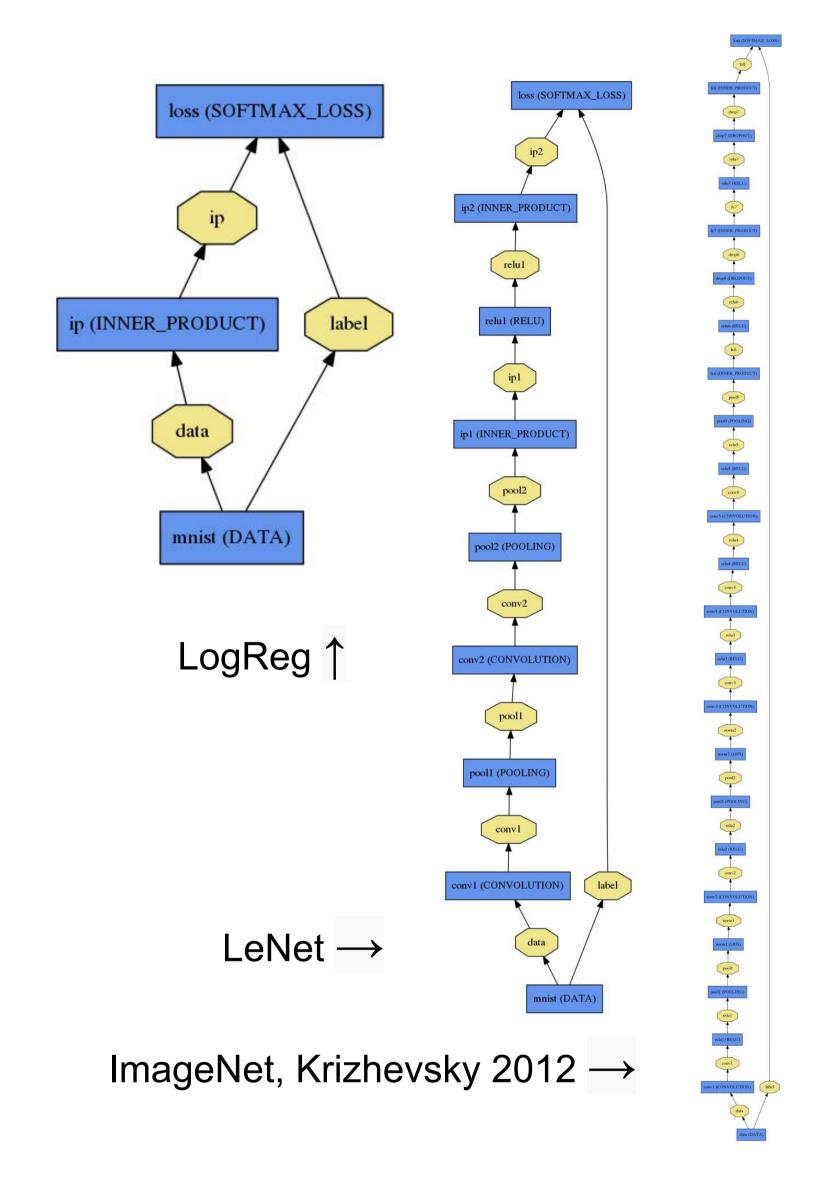
The Unreasonable Effectiveness of Deep Features



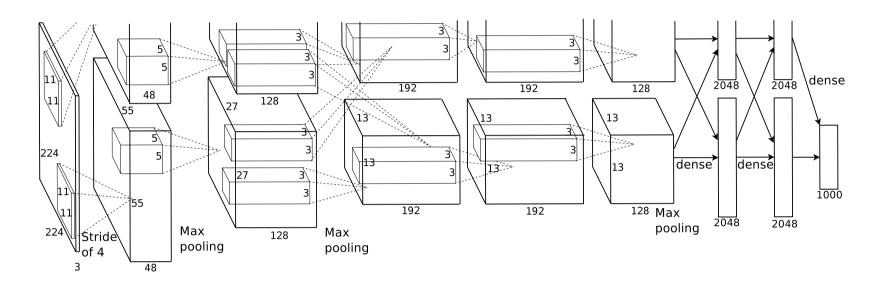
Feature Embedding maps visual features into a discriminative space.

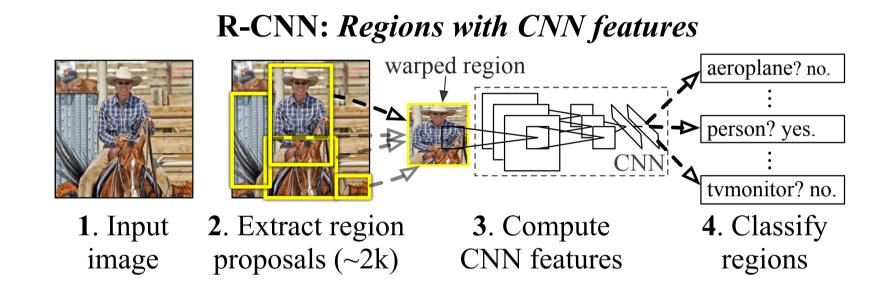
Classes separate in the deep representation!

Models



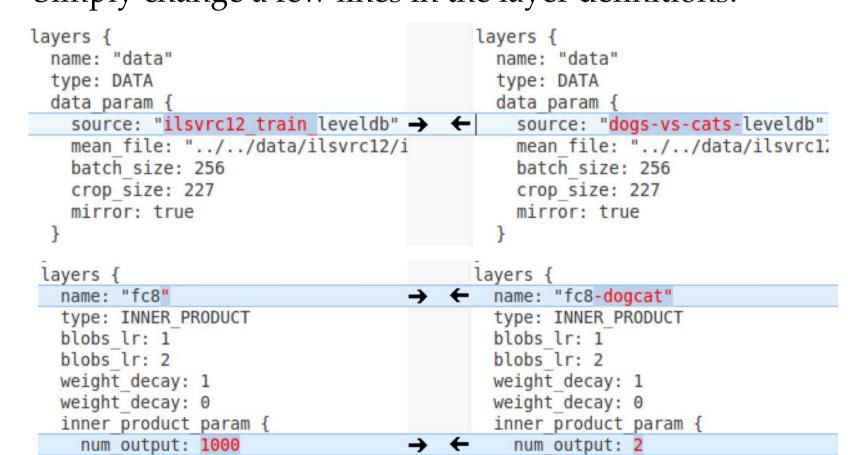
AlexNet: ImageNet Classification





Finetune Your Own Model

Simply change a few lines in the layer definitions.



Input: difference source

Output: different classifier

Seamless CPU / GPU Switch







Deployment

All with the same code.

References

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Acknowledgements

We would like to thank NVIDIA for GPU donation, the BVLC sponsors (http://bvlc.eecs.berkeley.edu/), and our open source community.

This material is based upon work supported by the National Science Foundation Graduate Research Fellowship under Grant No. DGE 1106400







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