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Training vision models with public transportation datasets

Computer Vision

Convolutional Neural Networks

Object Detection

Image Segmentation

Agenda

- Image Segmentation
- Fully Convolutional Networks (FCN)
- Training Vision Models Challenges

Image Segmentation

- Evolution from Classification → Detection → Segmentation
- Goal: Pixel level identification
- Metrics: Jaccard Index

$$- IoU = \frac{True\ Positive}{True\ Positive + False\ Positive + False\ Negative}$$

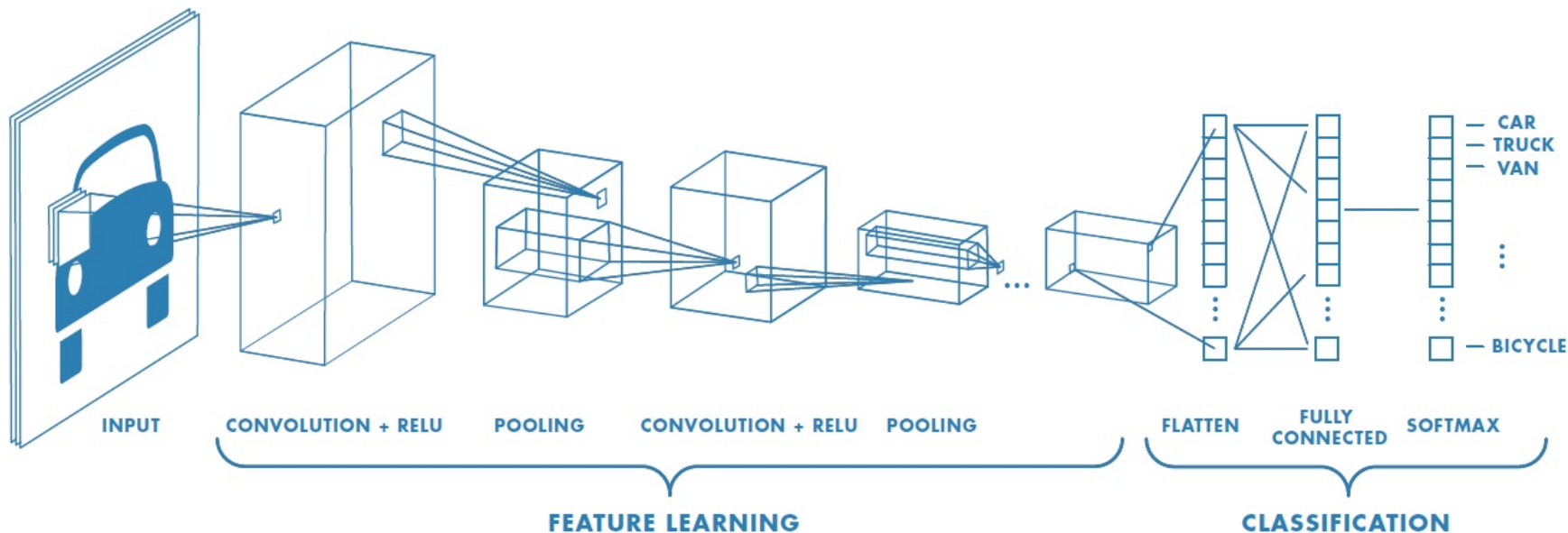
- Datasets:
 - PASCAL VOC
 - MS COCO
 - Cityscapes
 - ADE20K
 - Imagenet
 - KITTI

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Fully Convolutional Networks

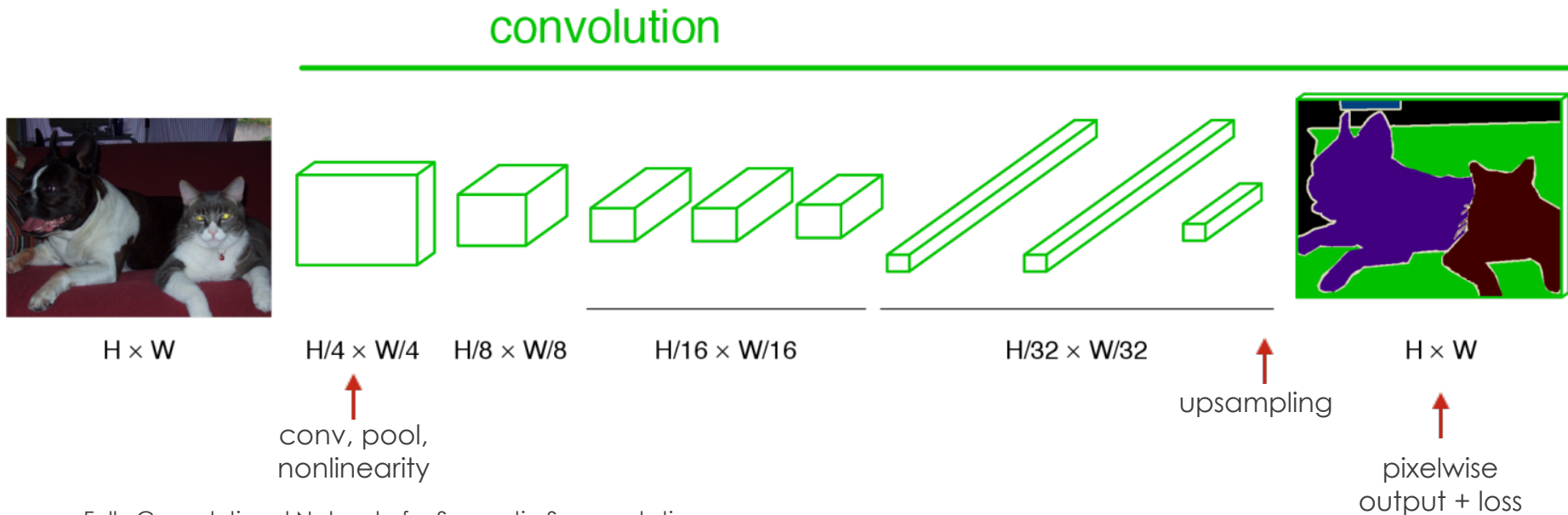
- Image Classification Objective: Predict Image Label



Source: Mathworks

Fully Convolutional Networks

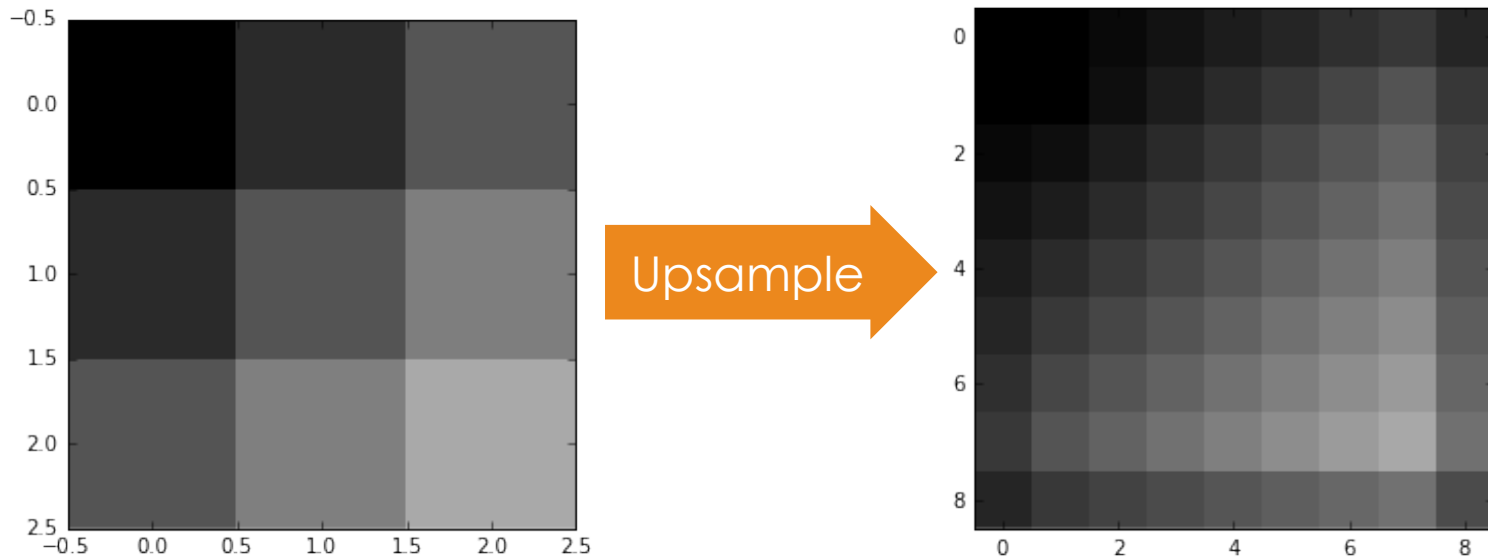
- Segmentation Objective: Predict Pixel Label



Fully Convolutional Networks for Semantic Segmentation

Evan Shelhamer Jonathan Long Trevor Darrell - UC Berkeley in CVPR'15, PAMI'16

FCN: Upsampling



Upsampling and Image Segmentation with Tensorflow and TF-Slim

Source: <http://warspringwinds.github.io/tensorflow/tf-slim/2016/11/22/upsampling-and-image-segmentation-with-tensorflow-and-tf-slim/>

Fully Convolutional Networks: Implementations

- Future easy to use– Coming to Keras: Dense Prediction API Design, Including Segmentation and Fully Convolutional Networks
 - <https://github.com/fchollet/keras/issues/6538#issuecomment-301342345>
- Original Caffe: <https://github.com/shelhamer/fcn.berkeleyvision.org>
- TensorFlow: <https://github.com/MarvinTeichmann/tensorflow-fcn>

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Training Vision Models Challenges: Time

Batch Size	epochs	Top-1 Accuracy	hardware	cost (\$)	time
256	100	58.7%	8-core CPU + K20 GPU	3,000	144h
512	100	58.8%	1 DGX station	129,000	6h 10m
4096	100	58.4%	1 DGX station	129,000	2h 19m
32K	100	58.5%	512 KNLs	1.2 million	24m

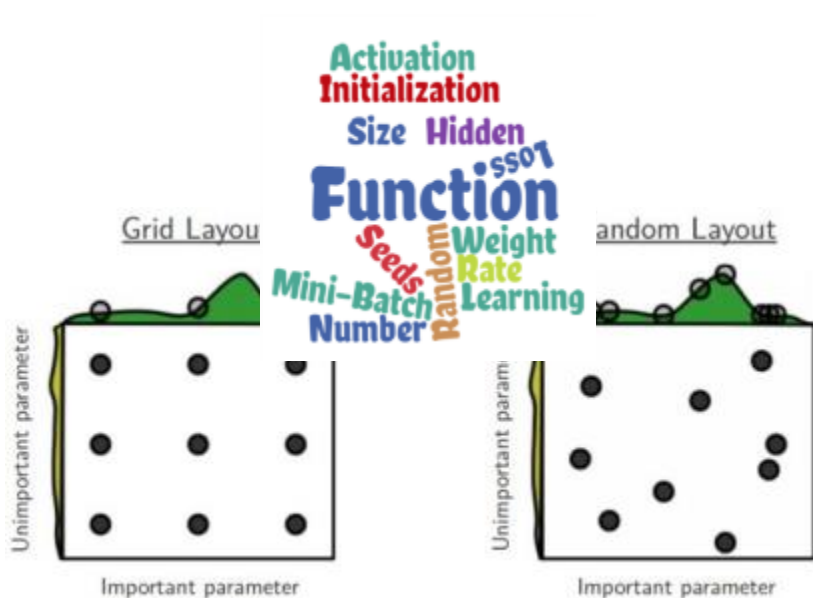
Table 7: The speed and hardware cost for training AlexNet.

For batch size=32K, we changed local response norm in AlexNet to batch norm.

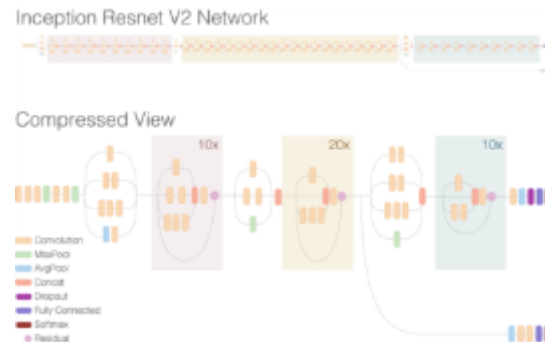
Batch Size	epochs	Top-1 Accuracy	hardware	cost (\$)	time
256	90	73.0%	1 DGX station	129,000	21h
8192	90	72.7%	1 DGX station	129,000	21h
8192	90	72.7%	32 DGX stations	4.1 million	1h
32K	90	72.4%	512 KNLs	1.2 million	1h

Table 8: The speed and hardware cost for training ResNet50. We did not use data augmentation.

Architectures & Hyper-Parameters

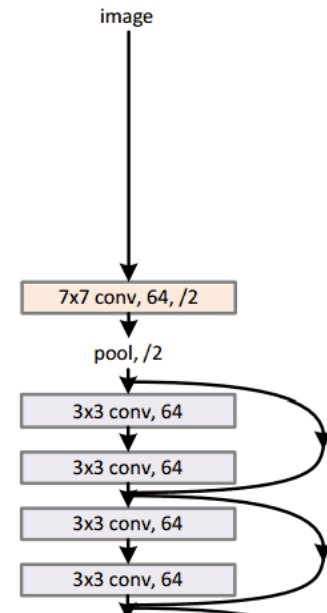


Hyper-Parameter Optimization



Google Inception v3

34-layer residual



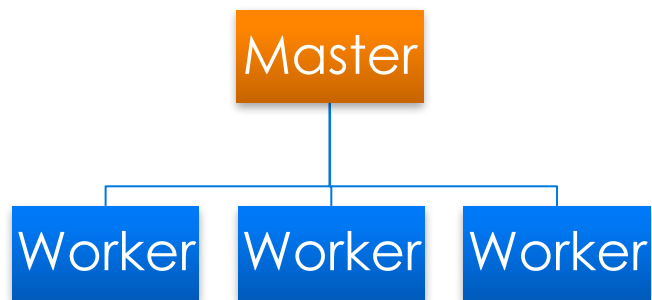
Microsoft Resnet

Architecture GIFs: <https://adeshpande3.github.io/adeshpande3.github.io/The-9-Deep-Learning-Papers-You-Need-To-Know-About.html>
More about hyper parameteric optimization http://colinraffel.com/wiki/neural_network_hyperparameters

Distributed Training

Data Parallel

- Distribute Data
 - Synchronous
 - Asynchronous



Model Parallel

- Distribute Model Operations

