Scaled Machine Learning at Matroid

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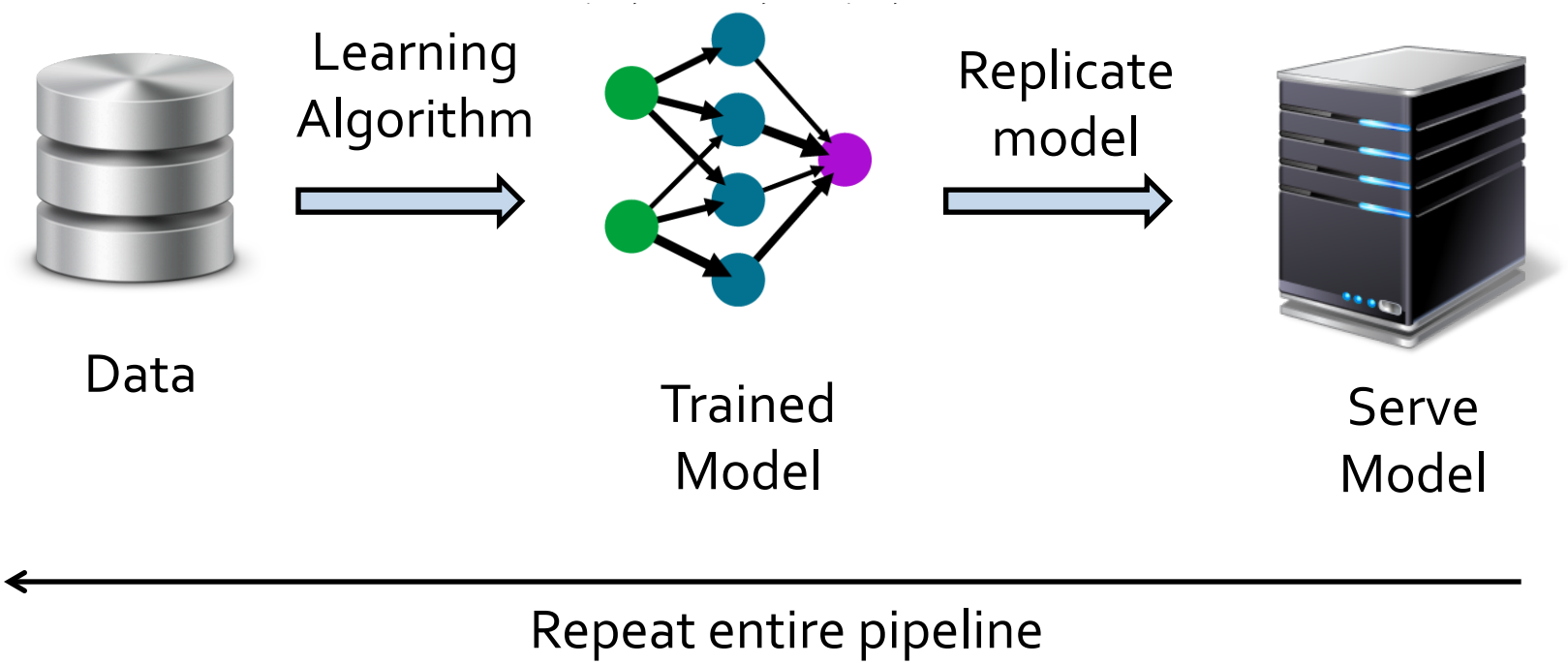
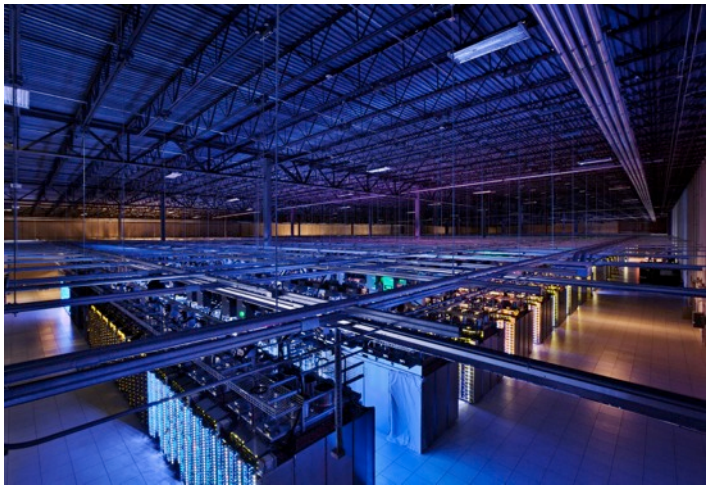
Machine Learning Pipeline



Scaling Machine Learning

Datasets and models growing faster than processing speeds

Solution is to parallelize on clusters and GPUs



Scaled ML at Matroid

Object recognition in Princeton ModelNet

» First on leaderboard for 40-class dataset

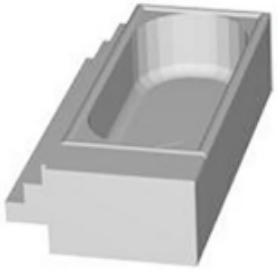
Matrix Computations and Optimization in Apache Spark

» Won KDD Best Paper Award runner-up

From Image Recognition to Object Recognition

Object recognition

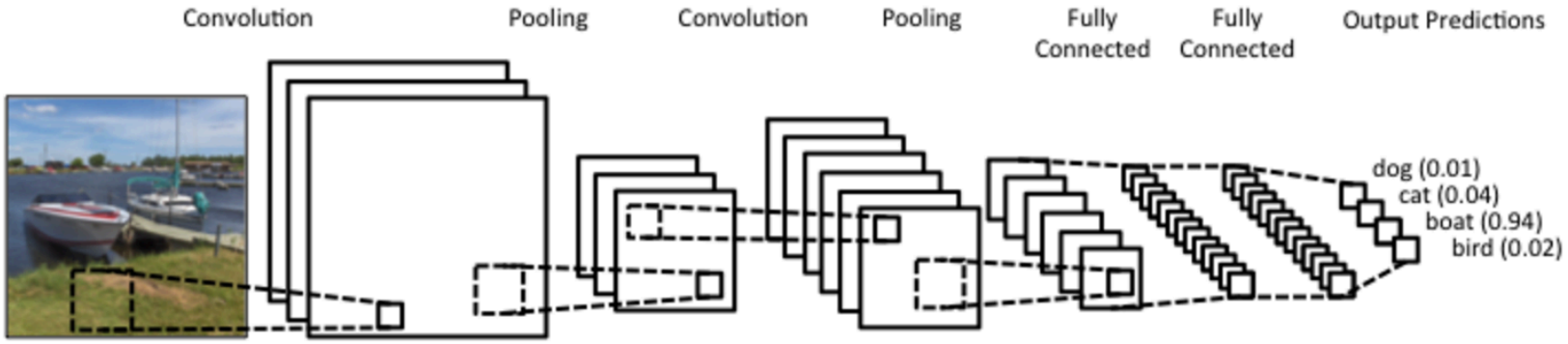
Given 3D model, figure out what it is



» bathtub

Try using image recognition on projections, but that only goes so far.

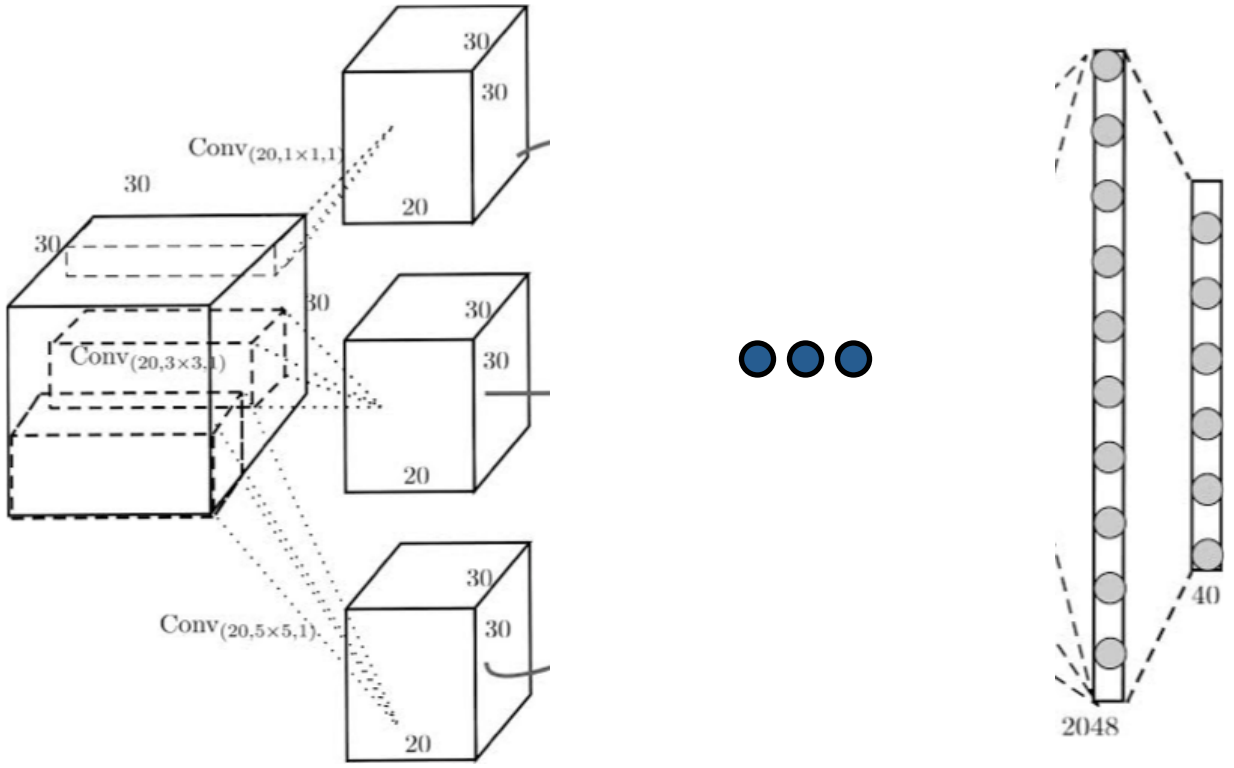
Convolutional Network



Slide a two-dimensional patch over *pixels*.

How to adapt to three dimensions?

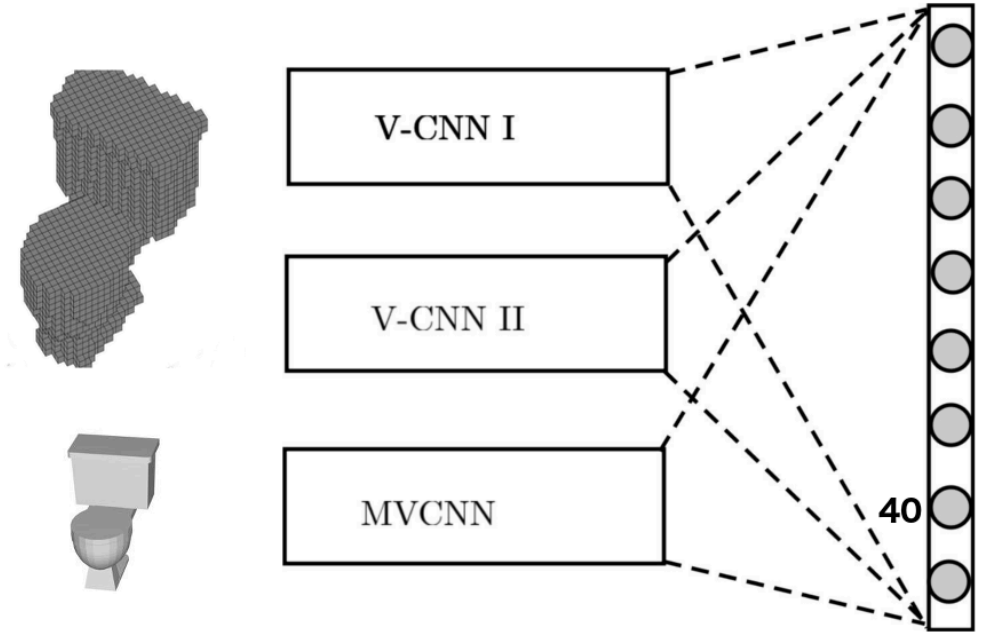
Volumetric (V-CNN)



Simple idea: slide a three-dimensional volume over *voxels*.

FusionNet

Fusion of two volumetric representation CNNs and one pixel representation CNN



Hyper-parameters tuned on a cluster

http://arxiv.org/abs/1607.05695

Matrix Computations and Optimization in Apache Spark

Traditional Network Programming

Message-passing between nodes (e.g. MPI)

Very difficult to do at scale:

- How to split problem across nodes?
  - Must consider network & data locality
- How to deal with failures? (inevitable at scale)
- Even worse: stragglers (node not failed, but slow)
- Ethernet networking not fast
- Have to write programs for each machine

Rarely used in commodity datacenters

Data Flow Models

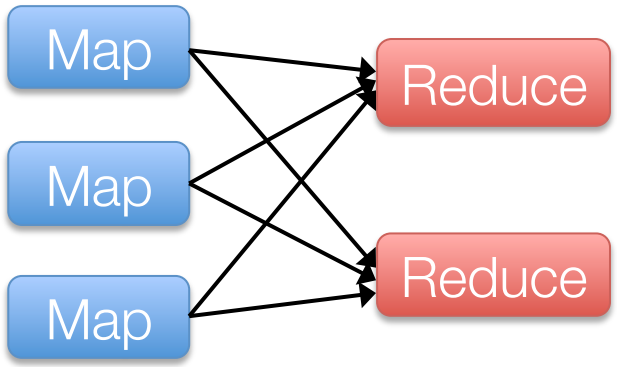
Restrict the programming interface so that the system can do more automatically

Express jobs as graphs of high-level operators

- » System picks how to split each operator into tasks and where to run each task
- » Run parts twice fault recovery

Biggest example: MapReduce

Nowadays: Spark, TensorFlow



Spark Computing Engine

Extends a programming language with a distributed collection data-structure

- » “Resilient distributed datasets” (RDD)

Open source at Apache

- » Most active community in big data, with 100+ companies contributing

Clean APIs in Java, Scala, Python, R

MLlib: Available algorithms

classification: logistic regression, linear SVM, naïve Bayes, least squares, classification tree, neural networks

regression: generalized linear models (GLMs), regression tree

collaborative filtering: alternating least squares (ALS), non-negative matrix factorization (NMF)

clustering: k-means||

decomposition: SVD, PCA

optimization: stochastic gradient descent, L-BFGS

Simple Observation

Matrices are often quadratically larger than vectors

A: n x n (matrix) $O(n^2)$

v: n x 1 (vector) O(n)

Even n = 1 million makes cluster useful

Spark TFOCS

Conic optimization program solver

Solve e.g. LASSO

$$\frac{1}{2}\|Ax-b\|_2^2+\lambda\|x\|_1$$

General Linear Programs

$$\text{minimize } c\cdot x+\frac{1}{2}\mu\|x-x_0\|_2^2 \text{ s.t. } Ax=b \text{ and } x\geq 0$$

Spark TFOCS

The implementation of TFOCS for Spark closely follows that of the Matlab TFOCS package.

Matrix Computations shipped to cluster, vector operations on driver

Come to KDD 2016 to learn more

Singular Value Decomposition

ARPACK: Very mature Fortran77 package for computing eigenvalue decompositions

JNI interface available via netlib-java

Distributed using Spark

Square SVD via ARPACK

Only interfaces with distributed matrix via matrix-vector multiplies

$$K_n = \begin{bmatrix} b & Ab & A^2b & \cdots & A^{n-1}b \end{bmatrix}$$

The result of matrix-vector multiply is small.

The multiplication can be distributed.

Thank you!

Matrix Computations paper

http://stanford.edu/~rezab/papers/linalg.pdf

FusionNet Object Recognition paper

http://arxiv.org/abs/1607.05695

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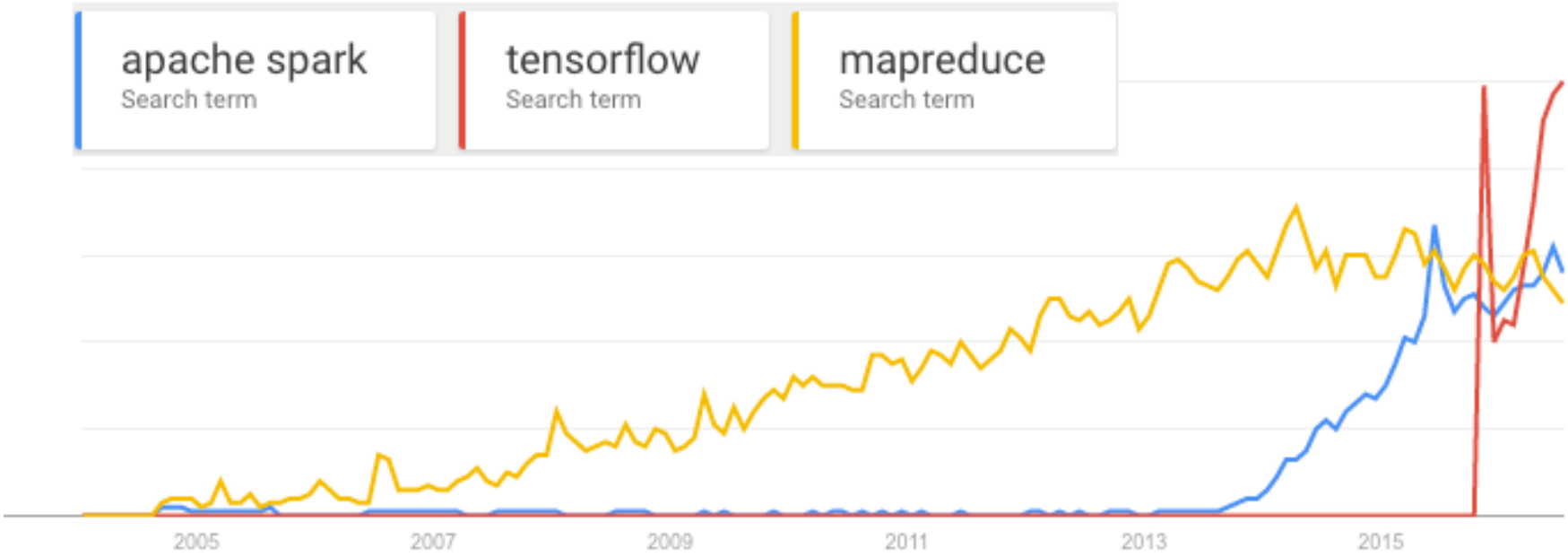
Apples and Oranges?



Source: google trends