



Systems Engineer

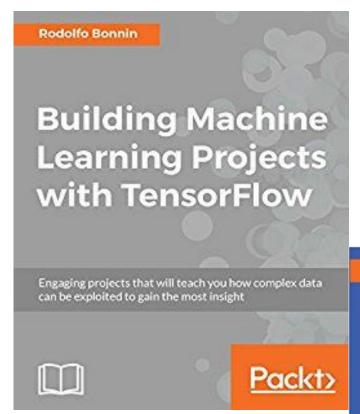
Ph.D Student, UTN

Working on Deep Learning since 2008

Ex ML @ Mercadolibre

CV Specialist @ Machinalis

Author of "Building Machine Learning Projects with Tensorflow" and "Machine Learning for Developers"











Visual Search Automatic Categorization Automatic Product Tagging

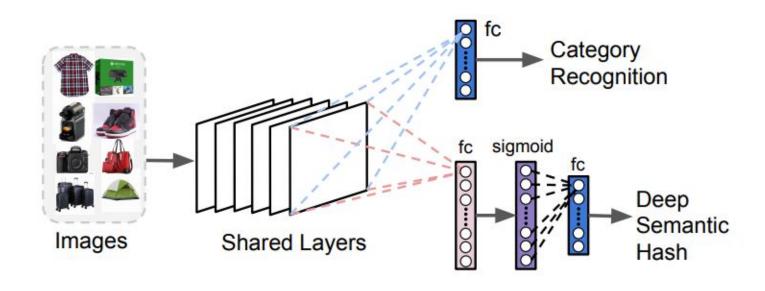
Moderation





> Visual detection & Search @ Ebay



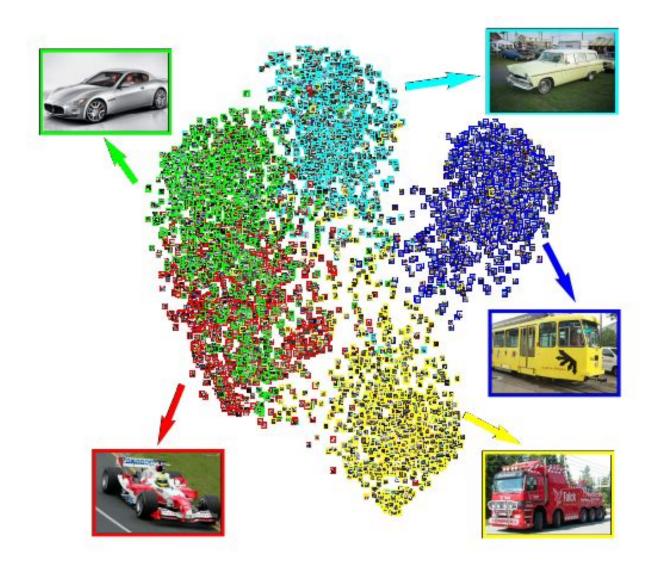


- Represent images as binary signatures instead of real values: reduces storage requirement and computation overhead.
- Hamming Distance

- DNN (ResNet50)
- Search only among top predicted categories and then use semantic binary hash with Hamming distance for fast ranking.
- For speed and low memory footprint, shared topology for both category prediction and binary hash extraction

> Visual detection & Search @ Ebay

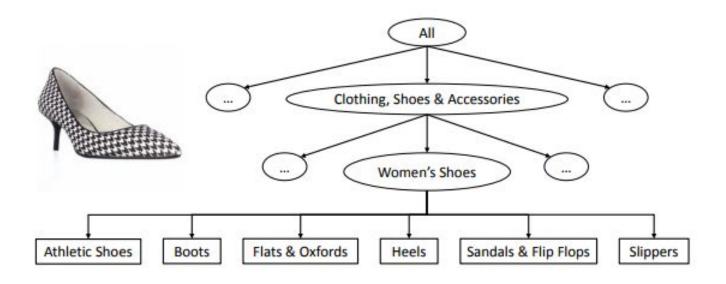




Experiments over imagenet:
T-SNE Over 4096 semantic features (non-parametric mapping!)

> Visual detection & Search @ Ebay





Item Specifics

Condition: New with box

Brand: Michael Kors

Heel Type:

Kitten

Style:

Kitten Heels

Heel Height:

Med (1 % in. to 2 3/4 in.)

US Shoe Size (Women's):

Material:

Haircalf

Width:

Medium (B, M)

Pattern:

Houndstooth

Color:

Multi-Colored

7

Aspects (Individual classifiers)



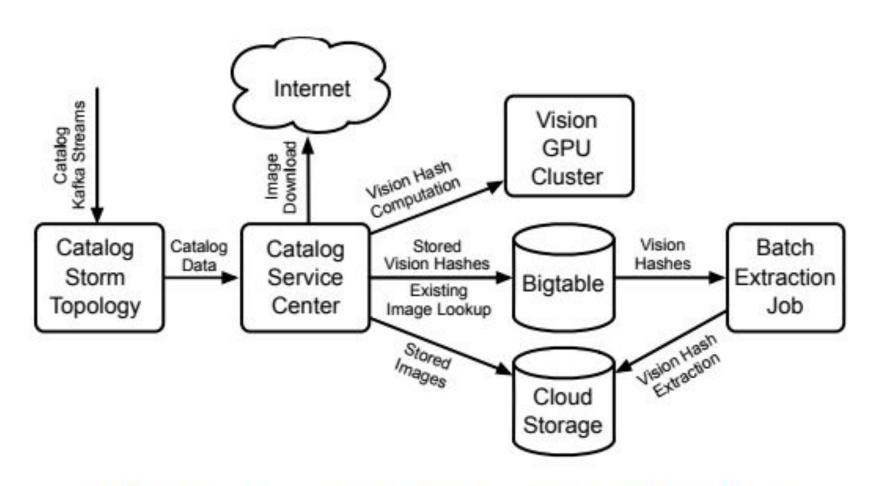


Figure 6: Image ingestion system architecture

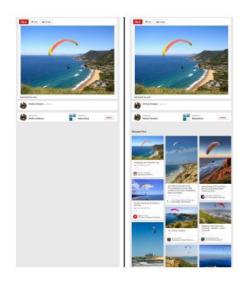




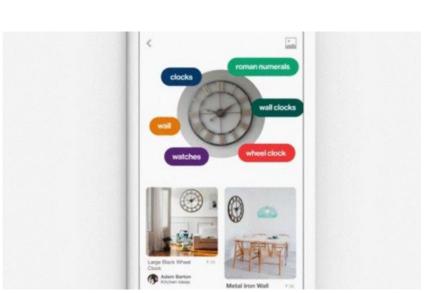
> Visual detection & Search @ Pinterest



- Related Pins (2014),
- Similar Looks (2015),
- Flashlight (2016)
- Lens (2017)







> Image representation and features



- Extraction of a variety of features from images, including local features and "deep features" extracted from the activation of intermediate layers of deep convolutional neural networks.
- Architectures based on AlexNet and VGG, extracting feature representations from fc6 and fc8 layers [Running on GPU]
- These features are binarized for representation efficiency and compared using Hamming distance.
- Click Through Rate based on Visual Features

> Two step object detection @ Pinterest



... fashion, **bags** & shoes hand picked by MoMo | See more ...



Figure 3: Instead of running all object detectors on all images, we first predict the image categories using textual metadata, and then apply object detection modules specific to the predicted category.

Table 3: Object detection performance.

	Faster R-CNN		SSD	
	precision	recall	precision	recall
Fashion	0.449	0.474	0.473	0.387
Home decor	0.413	0.466	0.515	0.360
Vehicles	0.676	0.625	0.775	0.775
Overall	0.426	0.470	0.502	0.371
Latency	272 ms		59 ms	





> Specialized providers: Vue.ai by Madstreetden









> Border detection







> Personal data embedded in text





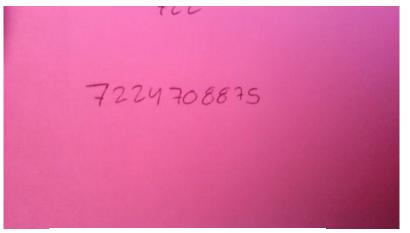
Recognized: ['.', '04142894437']



Recognized: ['0426-3905511', 'CHLOE']



["] [] ['o', 'SN', 'whats', '¡19433422485', 'EL', 'runa', '=', 'a', 'y']



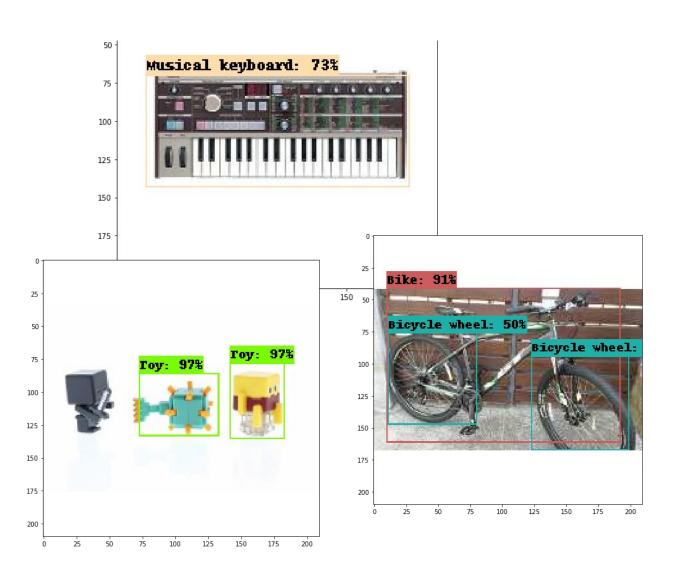
Recognized:['224208895']



['rd', '9351288065']

> Logo detection, category automatic detection





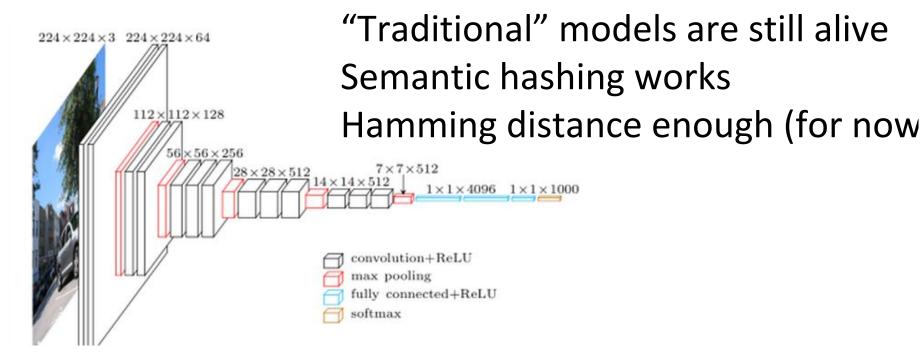
Score: 99.99998807907104

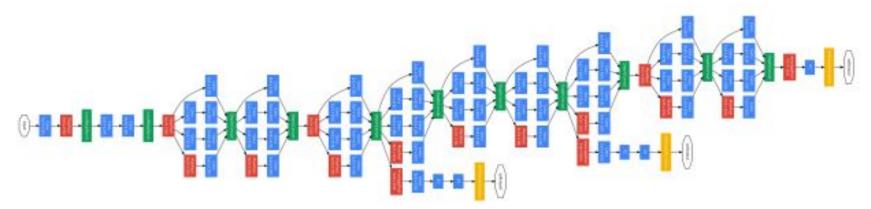


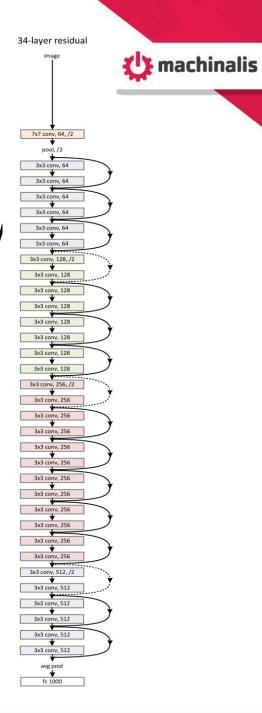




Summarizing









- Very specialized attributes detectors
- Jump to segmentation from detection
- Vertical marketplaces adopting ML
- Also generalized marketplaces (Shopify?)

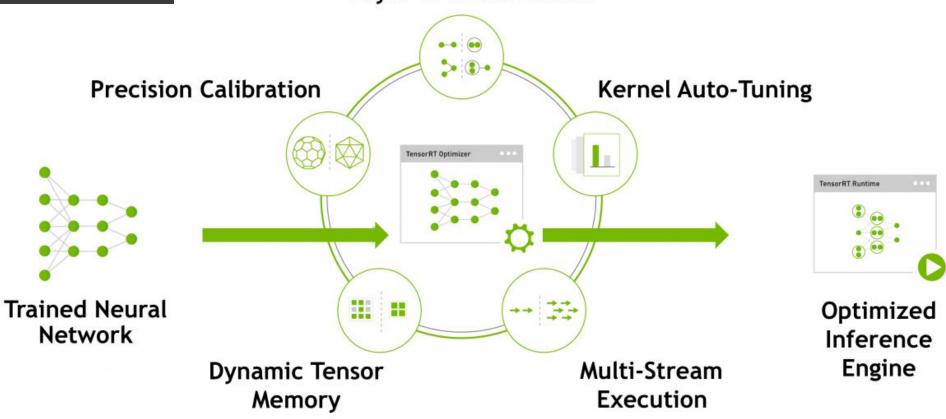




NVIDIA TensorRT

Programmable Inference Accelerator

Layer & Tensor Fusion



Numba & Friends



Caffe

TensorFlow ™

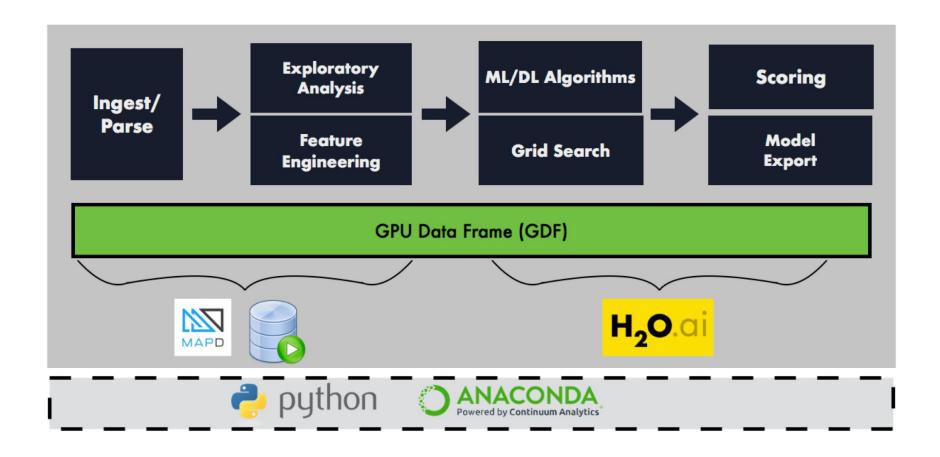


Keras



> Full GPU Analytics stack





Questions?

