



Connect w/ Data Science After Dark

1

Meetup

meetup.com/Data-Science-After-Dark

Upcoming Events + RSVP

2

YouTube

youtube.com/channel/UC7fA7eMv2745dleZ165pYMg

Live streams + past events

3

Slack

datascienceafterdark.slack.com

Join us in #data-science

4

Springfield Tech Calendar

fwdsgf.com

Local tech-focused events

Presented by: Jason Klein, Logic Forte @JasnK
Tuesday, April 21, 2020

meetup.com/Data-Science-After-Dark/events/268654605/

Who am I?

Jason Klein

I have been managing data since 2002. My background is a mix of IT, infosec, software. This led to my interest in security in Machine Learning.

[@JasnK](#)

<https://jrklein.com/>



Attacking a Machine Learning Model

Why we must **protect** Machine Learning models **critical** to our business?

Attackers who can access your model can manipulate inputs to achieve desired outputs. *e.g. Fraud Detection, Profanity and Image Filtering, etc.*



History of Machine Learning

1950 - Turing Test

1958 - Perceptron neural network

1967 - Nearest neighbor algorithm

1979 - The Stanford Cart navigates obstacles

1990s - Machine Learning shifts from knowledge-based to data-driven models

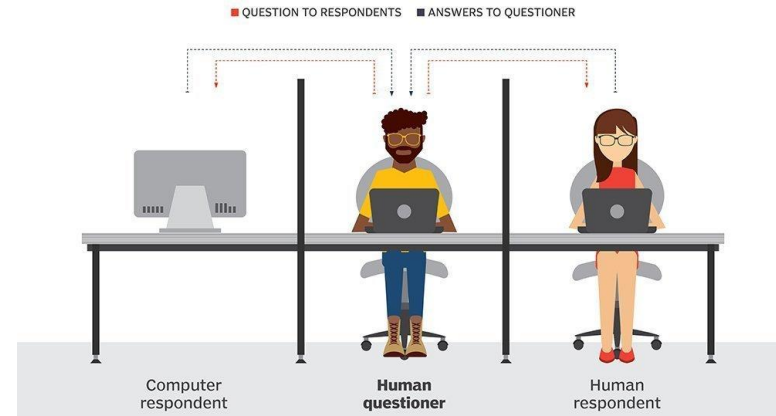
2006 - Deep Learning term coined

2010 - Microsoft Kinect tracks 20 features at 30 frames per second

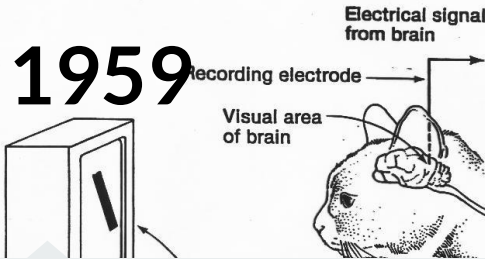
2010 ImageNet, 2011 Google Brain, 2012 YouTube Cat Finder, 2014 DeepFace

Turing test

During the Turing test, the human questioner asks a series of questions to both respondents. After the specified time, the questioner tries to decide which terminal is operated by the human respondent and which terminal is operated by the computer.

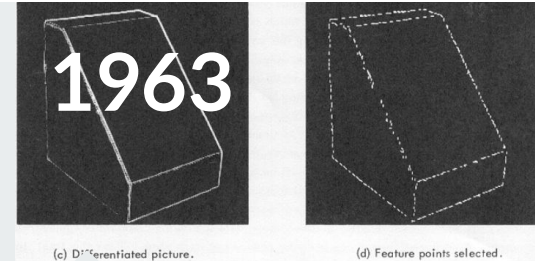


History of Image Classification



Hubel and Wiesel
discover core principal
of edge detection

Russell Kirsch
developed first digital
image scanner [\[1\]](#)[\[2\]](#)

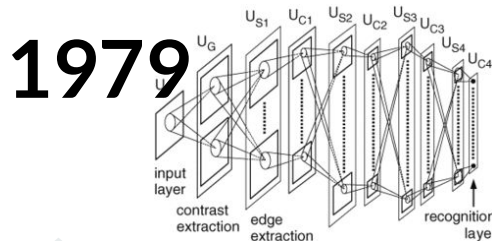


Lawrence Roberts
converts 2D photos
into line drawings



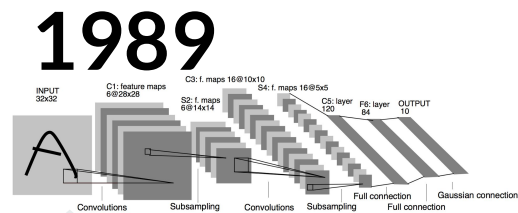
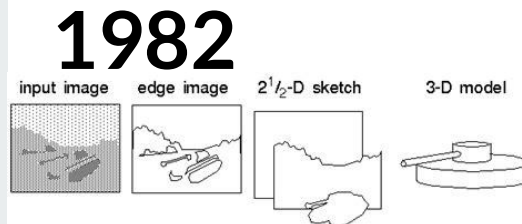
1959

History of Image Classification



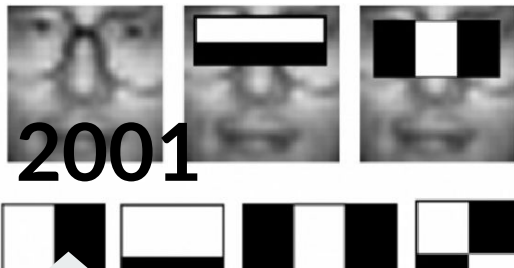
Kunihiko Fukushima
created neocognitron
neural network [\[1\]](#) [\[2\]](#)

David Marr
established that vision
is hierarchical [\[3\]](#)



Yann LeCun
LeNet-5, first modern
convnet, leading to
MNIST imageset [\[4\]](#) [\[5\]](#)

History of Image Classification

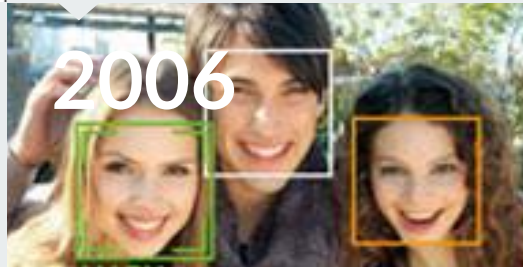


2001

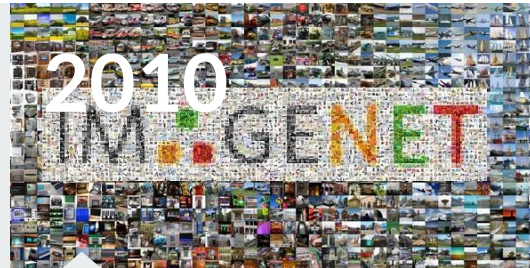
Viola and Jones
create real-time face
detection framework^[1]

Fujitsu

releases camera with
real-time Viola/Jones
face detection



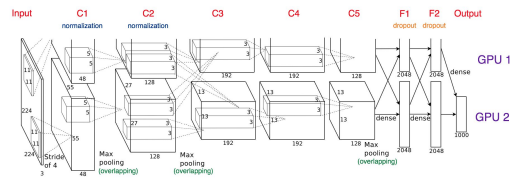
2006



2010

ImageNet project
begins annual contest,
recognition of 1000
image categories^{[2][3]}

History of Image Classification



2012

Alex Krizhevsky

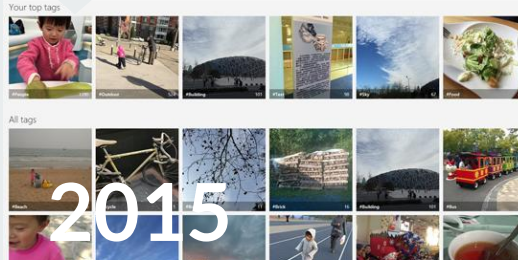
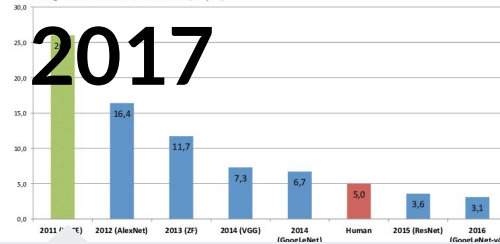
AlexNet wins ImageNet,
GPU model accuracy up
from 75% to 84.7% [\[1\]](#)[\[2\]](#)

Microsoft
team wins ImageNet
with 95.1% accuracy

[\[3\]](#)[\[4\]](#)

ImageNet Classification Error (Top 5)

2017



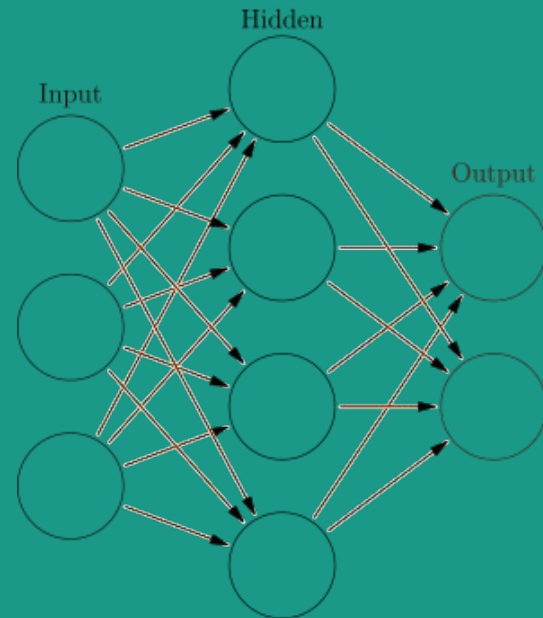
ImageNet

29 of 38 teams >95%
accuracy. Developing
3D competition. [\[5\]](#)

Basics of Image Classification

1- Train a Neural Network

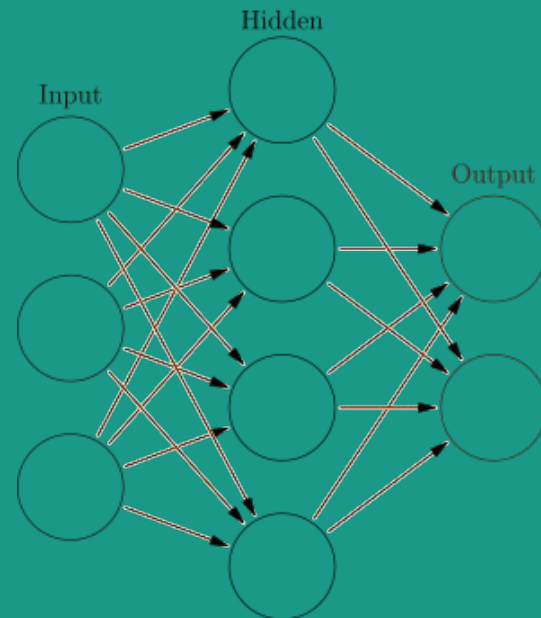
- Input Tagged Images
- Train Hidden Layers
- Output Neurons



Basics of Image Classification

2- Use the Neural Network

- Input Model File
- Input Image
- Output Prediction %

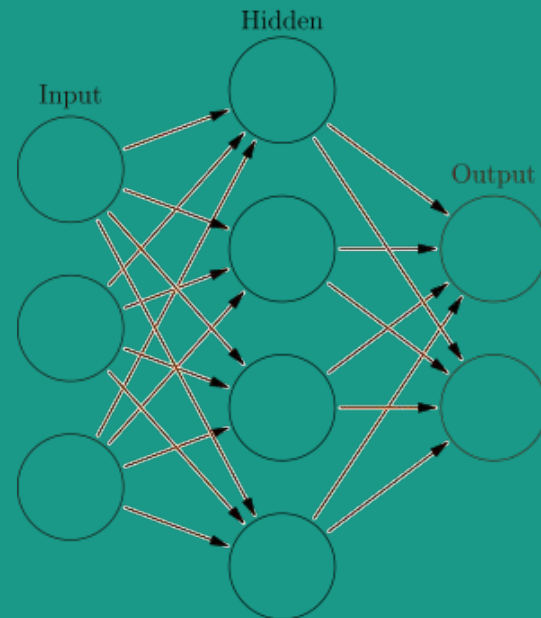


Basics of Image Classification

3- Profit

If you trained your own model, keep your model secret to help avoid attacks.

Ready to learn more? Deep Dive [\[1\]](#)[\[2\]](#)



Attacking a Machine Learning Model

I will demonstrate how easily we can attack an image classification model.

I will feed an image of a specific animal into the model and demonstrate how we can modify a single pixel in the original image to convince the model that the image is a different specific/desired animal.



Photo by Foo Bar

Dog or Frog .. 2

The goal here is to make an adversarial example using the photo of Trixi as a starting point. The website will classify it using the Keras pre-trained MobileNet network, for which you can download a copy. The image needs to do the following:

- Classify as a Tree Frog, at 95% confidence
- Be similar to the original image (max 2 bit difference using p hash)

This vulnerability is not specifically for MobileNet, and works against others as well. MobileNet was chosen as it uses less memory.



Dog or Frog .. 3

I want to make it clear that this isn't a stego, web, or "find a photo of the author's dog wearing a frog hat" problem. The intended solution is a photo that is clearly Trixi, but trick MobileNet into thinking there's a tree frog in it, rather than a dog.

A sample attack image is provided in the source code, that's recognized as a sealion. It looks squished due to the preprocessing of the network. It looks nearly identical to the preprocessed image without any attack present.



I can get 99+% confidence, and 0 bit difference.



Dog or Frog .. 5

```
Photo category      tree_frog
Photo is of a frog  True
Photo confidence     0.52180797
P hash distance from original photo  13
Top Preds
[('n01644373', 'tree_frog', 0.52180797),
 ('n01644900', 'tailed_frog', 0.4586374),
 ('n01675722', 'banded_gecko', 0.018022738),
 ('n01641577', 'bullfrog', 0.0010392488),
 ('n01694178', 'African_chameleon',
 0.0001692069)]
```



Dog or Frog .. 6

VERSIONS: macOS 10.15.4, python 3.7.0, pip 20.0.2

SOURCE: [solution.py](#)

```
$ cd ~/Code/dsad-picoctf-2018-dog-or-frog/  
$ pip install tensorflow keras Pillow numpy ImageHash  
$ python3 solution.py
```

Using TensorFlow backend.

```
Model's predicted likelihood that the image is a tree frog: 1.0298706e-09%  
Model's predicted likelihood that the image is a tree frog: 1.7842248%  
Model's predicted likelihood that the image is a tree frog: 87.817872%  
Model's predicted likelihood that the image is a tree frog: 98.974693%  
Model's predicted likelihood that the image is a tree frog: 90.498799%  
Model's predicted likelihood that the image is a tree frog: 99.204844%
```

```
$ ls -l  
-rw-r--r--@ 1 jrk staff 17271048 Apr 21 15:42 model.h5  
-rw-r--r-- 1 jrk staff 3856 Apr 21 15:54 solution.py  
-rw-r--r--@ 1 jrk staff 1641 Apr 21 15:49 solution_template.py  
-rw-r--r--@ 1 jrk staff 2534464 Sep 24 2018 trixi.png  
-rw-r--r-- 1 jrk staff 110468 Apr 21 16:17 trixi_frog.png  
-rw-r--r--@ 1 jrk staff 116308 Sep 24 2018 trixi_sealion.png
```



trixi_sealion.png trixi_frog.png



Summary

If you train ANY type of model for your organization, be aware that an attacker can use similar techniques to bypass your model if they can directly access your model.

For example, an attacker could feed a fraudulent transaction into a fraud detection model and determine what transaction detail can be changed to fool the model into believing the transaction is NOT fraudulent.

Data Science After Dark, Springfield Missouri

Tuesday, April 21, 2020

References

A Brief History of Computer Vision and Convolutional Neural Networks (2019)

hackernoon.com/a-brief-history-of-computer-vision-and-convolutional-neural-networks-8fe8aacc79f3

A Short History of Machine Learning -- Every Manager Should Read (2016)

forbes.com/sites/bernardmarr/2016/02/19/a-short-history-of-machine-learning-every-manager-should-read/

PicoCTF 2018: Dog or Frog Question

2018shell2.picoctf.com:11889

PicoCTF 2018 Writeup: General Skills (Dog or Frog Solution)

tcode2k16.github.io/blog/posts/picoctf-2018-writeup-general-skills/#dog-or-frog



Thank you for attending Data Science After Dark!

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