Image Classification

Special thanks to
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and
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Motivation

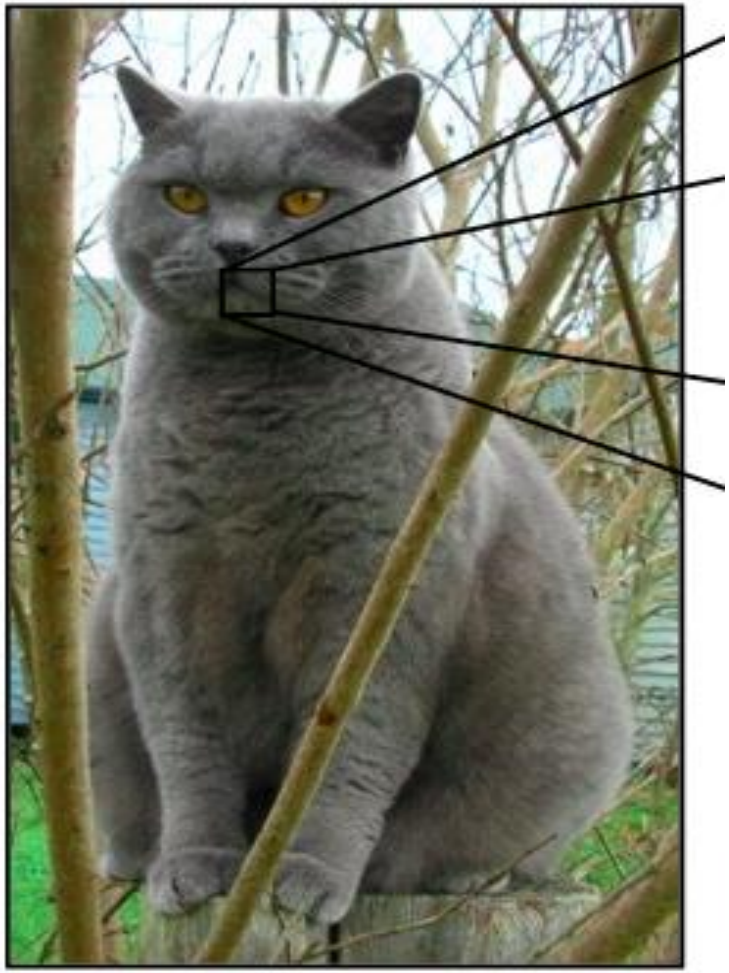
- Task: assign an input image one label from a fixed set of categories

- Why do you think it might be an important problem to solve?

Image classification example

Ignore the lines for now.

- image classification model takes an image and assigns probabilities to 4 labels *{cat, dog, hat, mug}*
- image is represented as one large 2-dimensional array of numbers + information about the colors
- image is 248 pixels wide, 400 pixels tall, and has three color channels **Red**, **Green**, **Blue**



http://cs231n.github.io/classification/

Image classification example

Image is 248 pixels wide, 400 pixels tall, and has three color channels **Red**, **Green**, **Blue**

The image consists of:

A: 248 + 400 + 3 numbers

B: 248 * 400 * 3 numbers

C: (248 * 400) ** 3 numbers

D: (248 + 400) ** 3 numbers

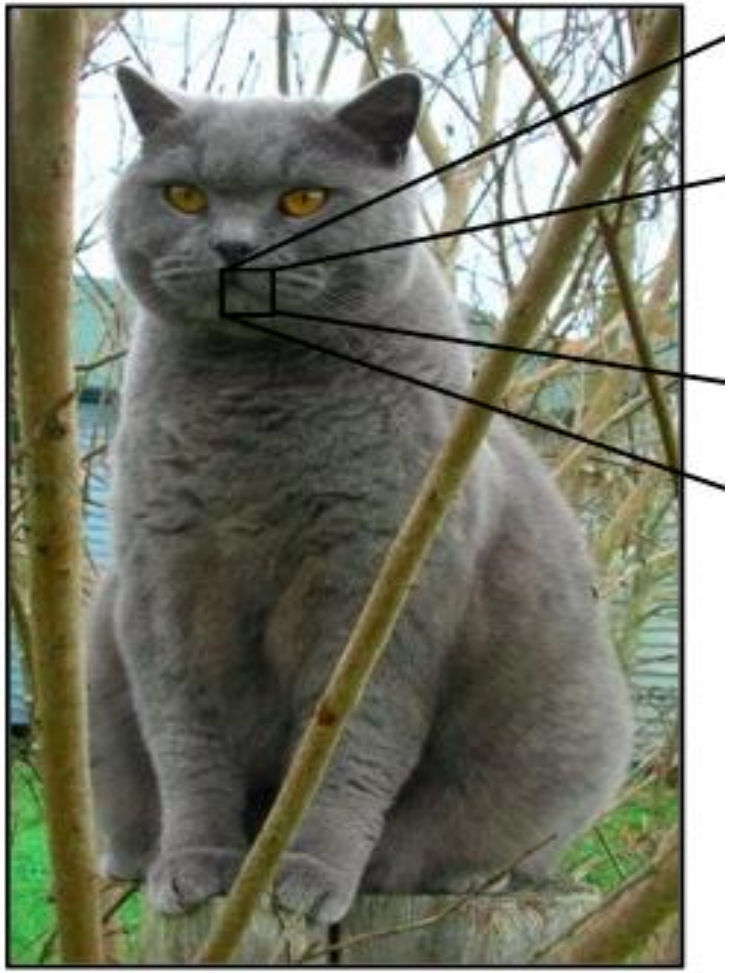
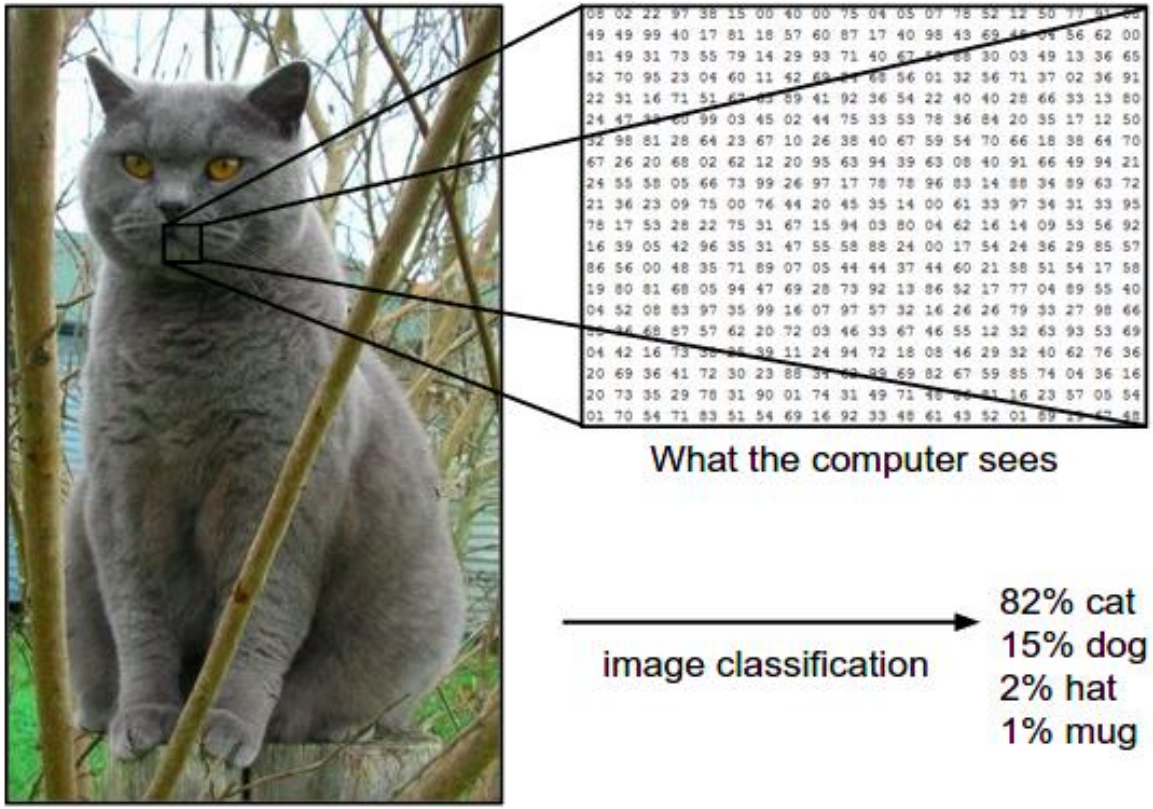


Image Classification Example



Challenges

Can you think of any challenges in classifying images?

(from a computer’s perspective)

Challenges

- **Viewpoint variation.**
- **Scale variation.**
- **Deformation**.
- **Occlusion.** The object of interest can be hidden. Sometimes only a small portion of an object (as little as few pixels) could be visible.
- **Illumination conditions.**
- **Background clutter**. The objects of interest may blend into their environment, making them hard to identify.
- **Intra-class variation.** The classes of interest can often be relatively broad.

http://cs231n.github.io/classification/

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Another related problem: recognizing handwritten digits

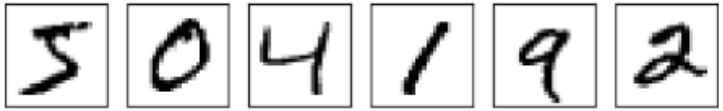
- Easy, right? How did you do it? Could you describe a “9” to me?

- But not for a computer.

- Application: sorting mail

Segmentation

- Convert 
- Then recognize an individual digit.



- It is worth solving the second problem first: recognizing individual digits.
- Once solved, segmentation problem can be solved as well:
  - One approach: try many ways to segment an image and score each trial
  - A trial segmentation gets a high score if the individual digit classifier is confident of its classification in all segments

Approach to classify individual digits

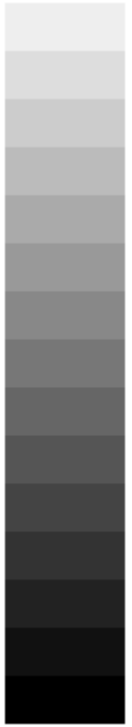
- take a large number of handwritten digits with labels. (Someone labeled them already).
- This is a training set:

Approach to solve it

- Develop a system which can *learn* from those training examples.
- By increasing the number of training examples, the network can learn more about handwriting, and so improve its accuracy.
    - Thousands, millions..

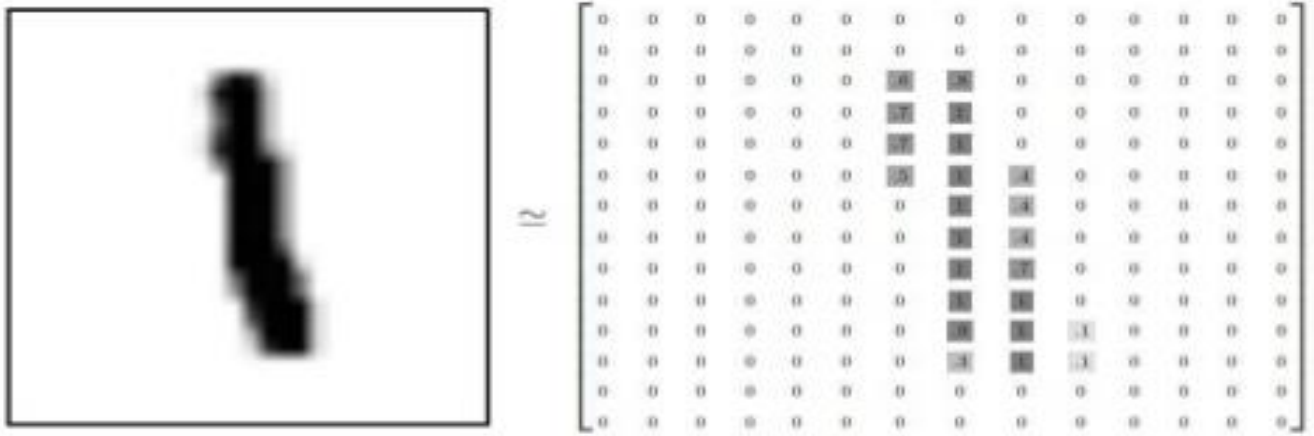
We have an image with a number. What is next?

- We need to convert an image to digits. How?
- Let’s say we have just black and white image.
- Now how about greyscale?



We have an image with a number. What's next?

MNIST Example (Predict the Number in the Image)



source: https://www.tensorflow.org/versions/master/tutorials/mnist/beginners/index.html

Training and Testing Sets

- MNIST (Modified National Institute of Standards and Technology) database
- Contains 60,000 training images and 10,000 testing images

Not every image can be placed in the database.

- Each image is 28 pixels by 28 pixels (total 784 numbers)
- Associated with a digit between 0-9.
- Anti - aliased (minimizing blocky patterns, when representing a high-resolution signal at a lower resolution, link to see an example)

https://en.wikipedia.org/wiki/MNIST_database

Artificial Neural Networks

- As a model of human brain

Neural networks can

- identify faces,
- recognize speech,
- read your handwriting,
- translate texts,
- play games (typically board games or card games)
- control autonomous vehicles and robots

https://appliedgo.net/perceptron/

Today's Exercise

Implement an algorithm that gives some evidence about whether an image has been photoshopped.

Details in Jupyter notebook in GitHub.