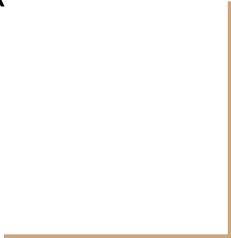




# Image Classification

Special thanks to  
Janine Tiefenbruck  
and  
Marina Langlois



# Motivation

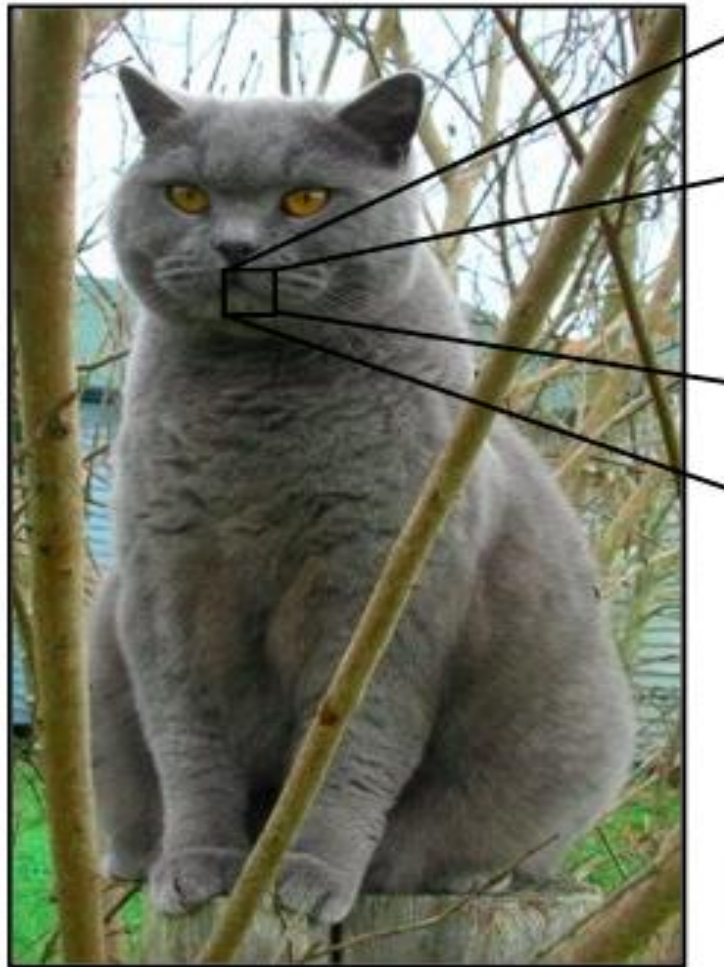
- Task: assign an input image one label from a fixed set of categories
- Why do you think it might be an important problem to solve?

# 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**



# Image classification example

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

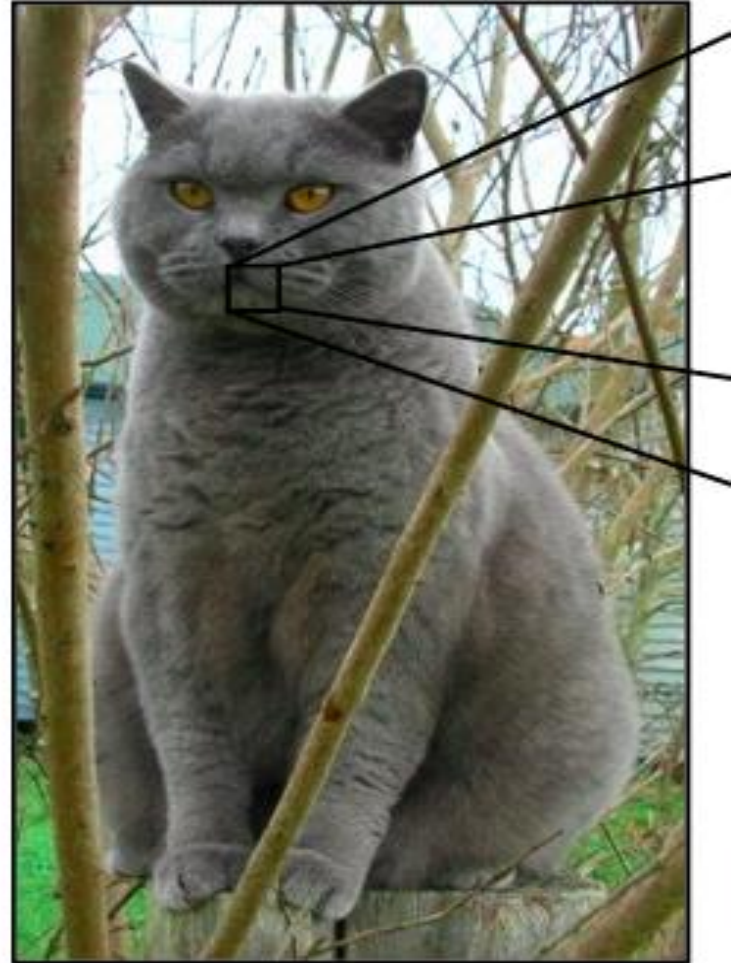
The image consists of:

A:  $248 + 400 + 3$  numbers

B:  $248 * 400 * 3$  numbers

C:  $(248 * 400) ** 3$  numbers

D:  $(248 + 400) ** 3$  numbers



# Image Classification Example



08	02	22	97	38	15	00	40	00	75	04	05	07	78	52	12	50	77	81	85
49	49	99	40	17	81	18	57	60	87	17	40	98	43	69	45	08	56	62	00
81	49	31	73	55	79	14	29	93	71	40	67	53	03	30	03	49	13	36	65
92	70	95	23	04	60	11	42	63	24	68	56	01	32	56	71	37	02	36	91
22	31	16	71	51	67	83	59	41	92	36	54	22	40	40	28	66	33	13	80
24	47	35	00	99	03	45	02	44	75	33	53	78	36	84	20	35	17	12	50
32	98	81	28	64	23	67	10	26	38	40	67	59	54	70	66	18	38	64	70
67	26	20	68	02	62	12	20	95	63	94	39	63	08	40	91	66	49	94	21
24	55	58	05	66	73	99	26	97	17	78	78	96	83	14	88	34	89	63	72
21	36	23	09	75	00	76	44	20	45	35	14	00	61	33	97	34	31	33	95
78	17	53	28	22	75	31	67	15	94	03	80	04	62	16	14	09	53	56	92
16	39	05	42	96	35	31	47	55	58	88	24	00	17	54	24	36	29	85	57
86	56	00	48	35	71	89	07	05	44	44	37	44	60	21	58	51	54	17	58
19	80	81	68	05	94	47	69	28	73	92	13	86	52	17	77	04	89	55	40
04	52	08	83	97	35	99	16	07	97	57	32	16	26	26	79	33	27	98	66
55	46	68	67	57	62	20	72	03	46	33	67	46	55	12	32	63	93	53	69
04	42	16	73	58	24	39	11	24	94	72	18	08	46	29	32	40	62	76	36
20	69	36	41	72	30	23	88	34	62	89	69	82	67	59	85	74	04	36	16
20	73	35	29	78	31	90	01	74	31	49	71	48	86	81	16	23	57	05	54
01	70	54	71	83	51	54	69	16	92	33	48	61	43	52	01	89	28	47	48

What the computer sees

image classification

82% cat  
15% dog  
2% hat  
1% mug

# 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.

# Guess

- Viewpoint variation.
- Scale variation.
- Deformation.
- Occlusion.
- Illumination conditions.
- Background clutter.
- Intra-class variation.





# Guess

- **Viewpoint variation.**
- **Scale variation.**
- **Deformation.**
- **Occlusion.**
- **Illumination conditions.**
- **Background clutter.**
- **Intra-class variation.**



# Guess

- **Viewpoint variation.**
- **Scale variation.**
- **Deformation.**
- **Occlusion.**
- **Illumination conditions.**
- **Background clutter.**
- **Intra-class variation.**



# Guess

- Viewpoint variation.
- Scale variation.
- Deformation.
- Occlusion.
- Illumination conditions.
- Background clutter.
- Intra-class variation.



# Another related problem: recognizing handwritten digits

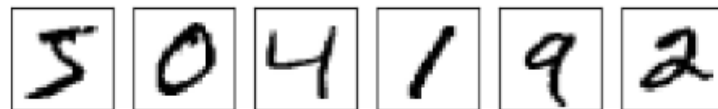
- Easy, right? How did you do it? Could you describe a “9” to me?

504192

- But not for a computer.
- Application: sorting mail

# Segmentation

- Convert *504192*
- 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?

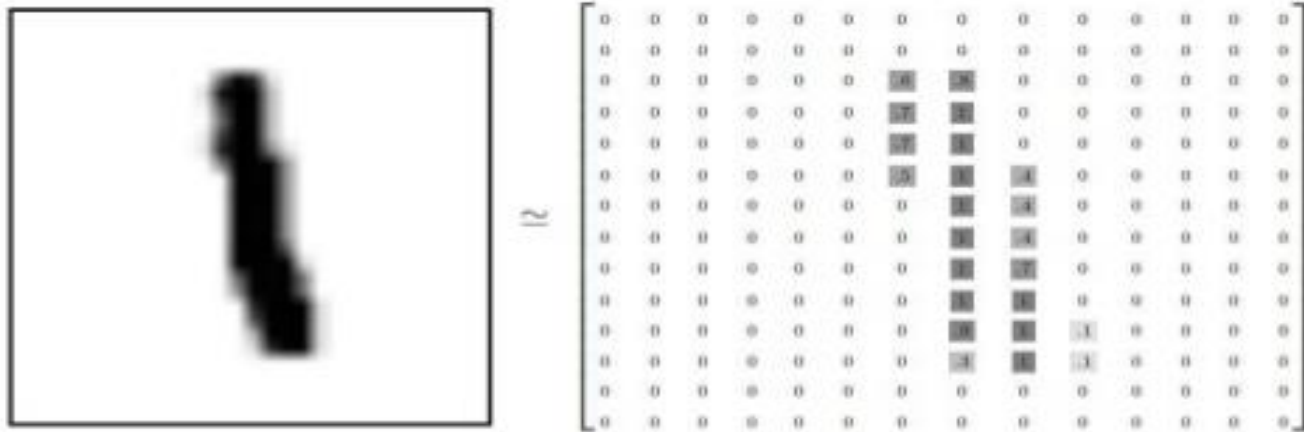
- We need to convert an image to digits. How?
- Let's say we have just black and white image.
- Now how about greyscale?





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

## MNIST Example (Predict the Number in the Image)



# 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](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

# Today's Exercise

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

Details in Jupyter notebook in GitHub.