#### CS 181AI Lecture 2

## Intro to ML Models

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#### Logistics

- Assignment 1 released this afternoon, due next **Monday, Jan 30** (one week turnaround because 90% of it will be done after today's demo)
- Code from class will be available this afternoon as well
- Reading group assignments will be sent by tomorrow morning, and first four papers will be on course webpage by then as well

#### Setup for Today

• Navigate to colab.research.google.com and create a new notebook

#### Today's Plan

- Background on machine learning models
- Creating a neural network vs. using it today's focus is on using it
- Get set up with downloading an existing vision model (that someone else created) and using it on our images

# Algorithms and Models and Neural Networks (oh my!)

- Algorithm is the pattern-recognition procedure that's run on data to create a model
- Neural network is just one type of machine learning model



#### Neural Networks



- Process of using data to create the model: called **training**
- For today, training process = black box

#### Neural Networks



- Process of using data to create the model: called training
- For today, training process = black box
- Then how do we get a usable model??

#### **Pre-trained Models**

- Other people have trained on large public datasets to create models that you can use "off the shelf"
- For vision models, they can only identify specific objects

#### Widely Used Datasets

- ImageNet
- Pascal
- MNIST
- CIFAR



#### Structure of a Model

- Model consists of layers
- We'll focus mostly on vision models, in particular convolutional neural networks (CNNs)



#### Roles of Different Layers

• Early layers **extract features**, last layers **classify image** 



- Navigate to colab.research.google.com and create a new notebook
- Add the following lines to the top.

import torch
import torchvision

#### Torch & Torchvision

- torch: open-source ML library used for creating neural networks
- torchvision: open-source computer vision library that contains popular datasets, model architectures, etc.
- We'll be using the "PyTorch framework"
  - PyTorch: open-source ML framework based on Python and torch

import torch
import torchvision

- Let's download an existing model. ResNet50 is a common vision model structure. It has 50 layers and is used to classify between different objects
- Weights identify which dataset was used for training (more on this later)
- In this case, ImageNet was used

resnet50\_model = torchvision.models.resnet50(weights=torchvision.models.ResNet50\_Weights.IMAGENET1K\_V1)

- Try printing the model
- What types of layers do you see?

### Types of Layers

- Convolutional
- Linear
- Other items you might see when printing a model
  - Relu
  - BatchNorm

#### Inputs of a Model

- Tensor -> matrix
- Each step is matrix operation
- Weights are the matrices we multiply the inputs by

- Set up image access:
  - Put image in Google drive (use the same Google account you're using to access colab)
  - Import Image from PIL (Python Imaging Library)
  - Mount the Google drive

from PIL import Image
from google.colab import drive
drive.mount('/content/drive')

- Set up image access
- Open the image:
  - In this example, I put golden.jpeg in a folder called colab\_resources in my Google drive and added the line below



- Set up image access
- Open the image
- Convert image to a tensor:
  - Torchvision has many types of transforms to manipulate the image before feeding it through the model. The ones that are important to us are:
    - Resize(x) -> resizes so that the aspect ratio stays the same and the height is x
    - CenterCrop -> crop to the center 224
    - ToTensor -> converts image to tensor
    - Normalize -> normalize with mean and std dev of the whole dataset

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  - .unsqueeze(0) -> adds another dimension to the front, e.g. (3, 224, 224) becomes (1, 3, 224, 224)

- Parsing the output:
  - Each value represents 1 class from ImageNet -> the model's prediction of how likely the image is to be that class
  - Softmax: rescale so that elements lie in the range [0,1] and sum to 1