

CS 181AI
Lecture 5

Training cont. + Object Detectors

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Logistics

- Assignment 2 is out

Today's Plan

- Group B leads reading discussion
- Finish discussion about debugging low accuracy when training a neural network
- Object Detection models

Course Path: The First Half

Intro to ML models and hands-on experience working with them

- Run a pre-trained model over your own images (1/23)
- Learn about training process and train a small model on a common dataset (1/25)
- Debug common issues with low accuracy when training (1/30)
- Learn about different types of models and their training processes (2/1)

Ethical AI

- Assess goals in building models
- Assess methods of data collection
- Study how various types of bias affect model results
- Study real use-cases

Computational Resources to Run and Train Models

- Learn about properties of machines that are particularly good for ML
- Study state-of-the-art ML devices + some promising future computing methods (e.g., quantum)
- Learn about how using multiple machines can enable scaling of ML tasks
- Study methods for lowering computational resources needed

Other Resources to Run and Train Models: Memory, Energy

- Compare memory and energy usage of running and training various types of models
- Study strategies for lowering memory and energy usage
- Case study: ML for monitoring agriculture

Improving Model Accuracy

- Common culprits:
 - Data
 - Model
 - Training Process

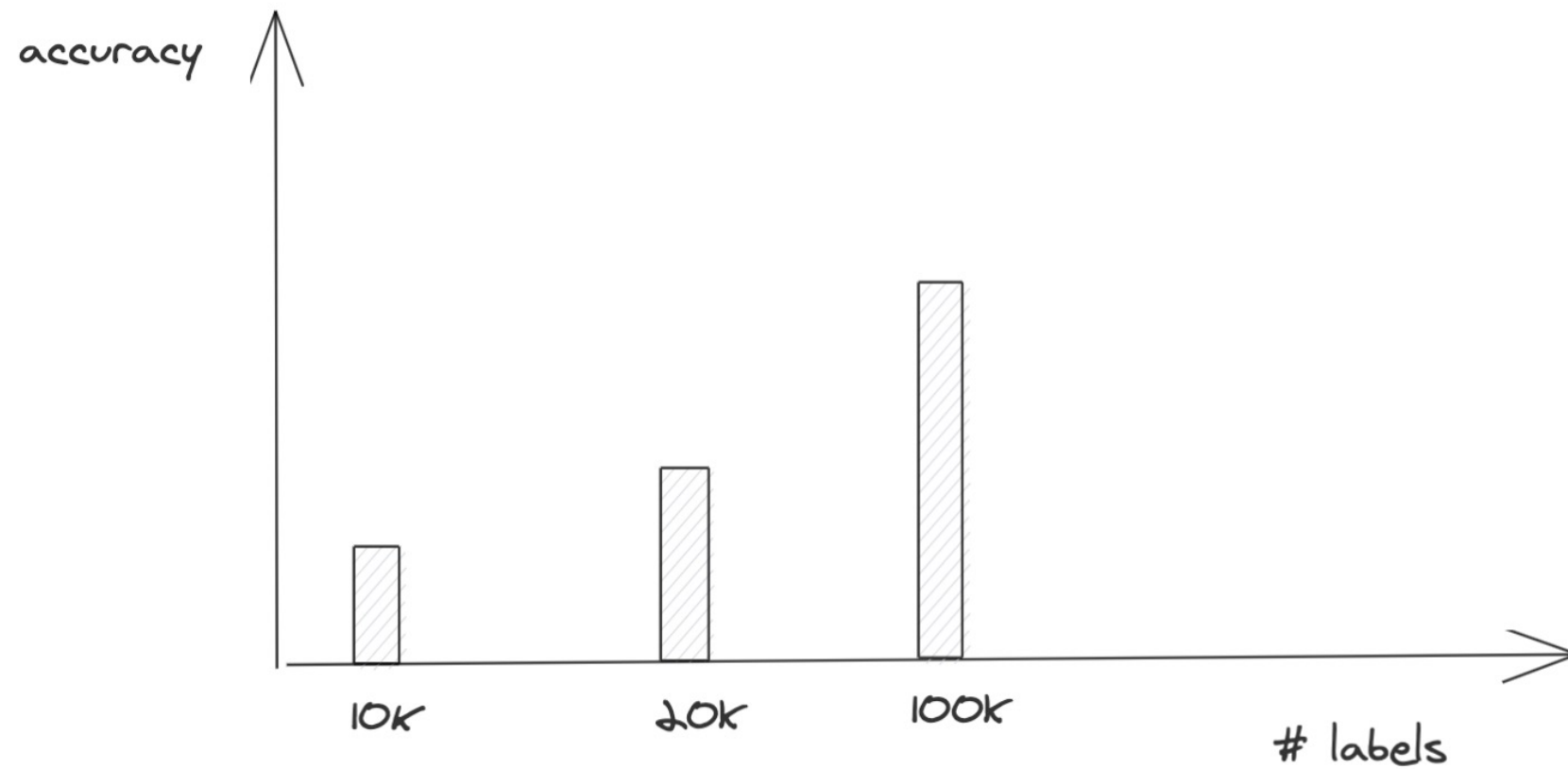
Labeling Example

- How many **entities** are in the following sentence?

Darth Sidious, known simply as the Emperor, was a Dark Lord of the Sith who reigned over the galaxy as Galactic Emperor of the First Galactic Empire.

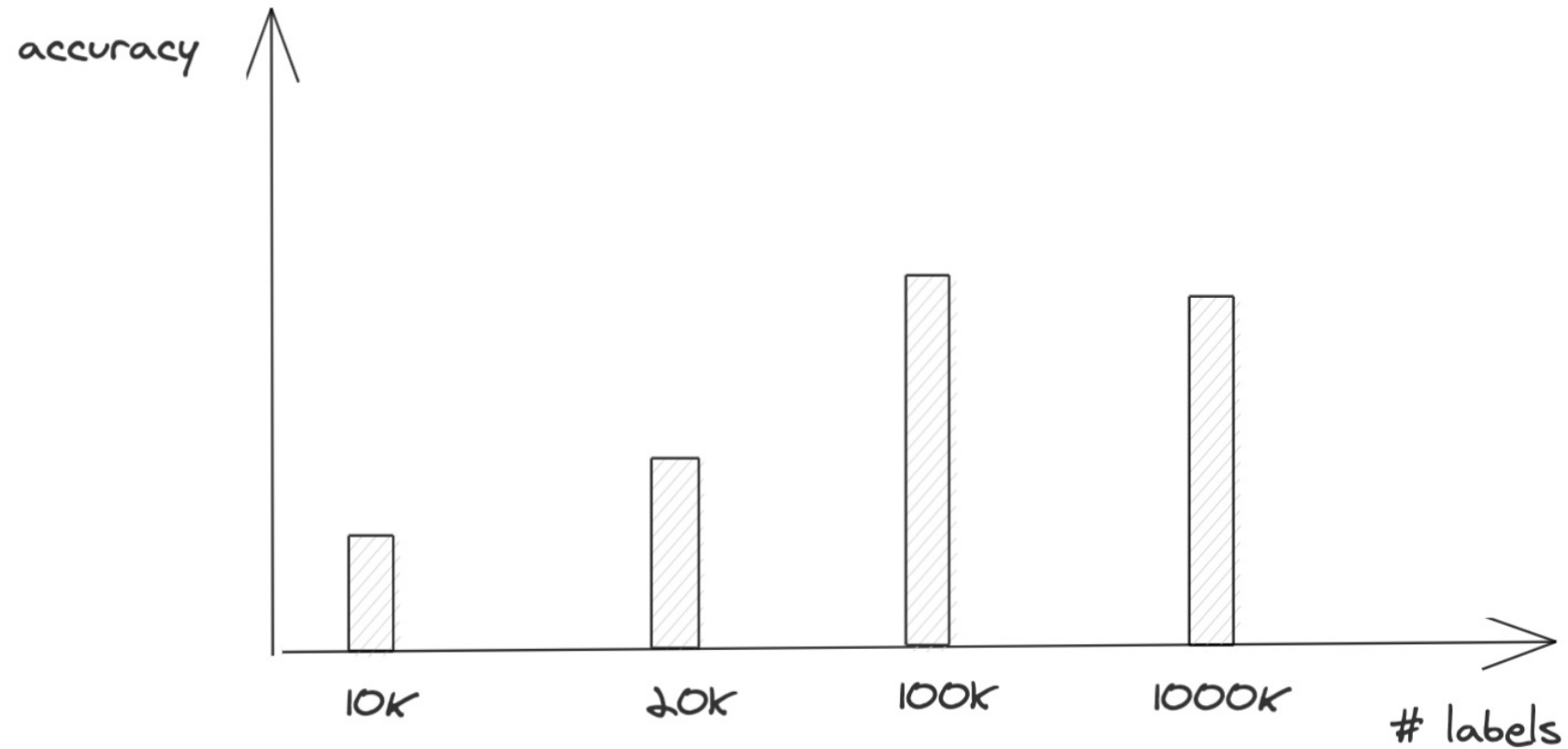
Labeling

- More data isn't always better



Labeling

- More data isn't always better



Labeling

- Programmatic Labeling can be messy but much faster and more adaptive
- Semi-supervision, or self-training: start with initial set of labels and train on new samples - only include the ones with a high score
- Active learning – only label those that model is uncertain about

Class Imbalance

- Which model do you hope your bank uses?

Model A	Actual Fraud	Actual Normal
Predicted Fraud	10	10
Predicted Normal	90	890

Model B	Actual Fraud	Actual Normal
Predicted Fraud	90	90
Predicted Normal	10	810

Class Imbalance

- Which model do you hope your bank uses?
 - Just going by accuracy, they both have an accuracy of 90%

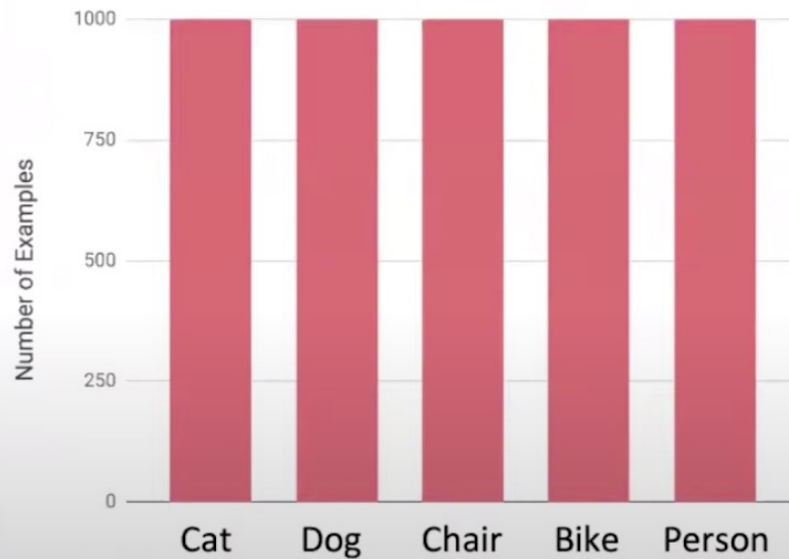
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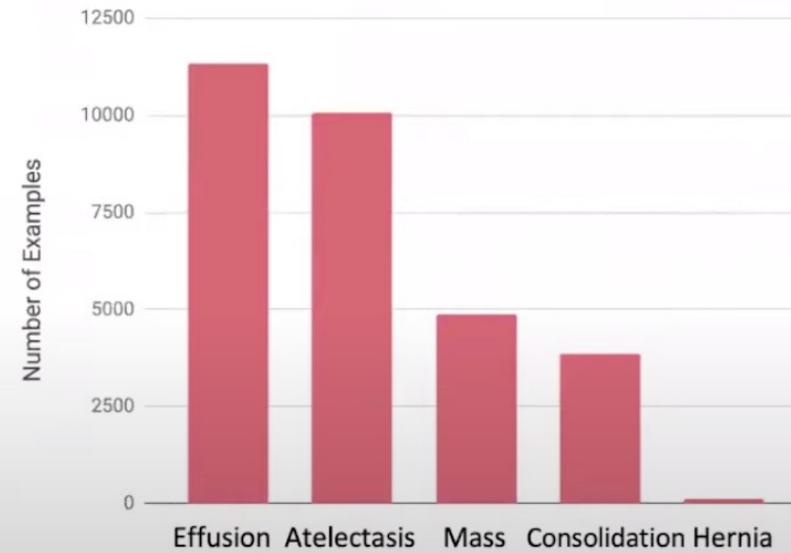
Class Imbalance

Small data and rare occurrences

ML works well when the data distribution is this:



Not so well when it is this:



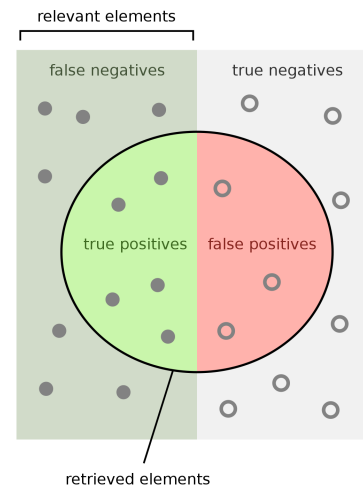
Class Imbalance

- Statistically, predicting the majority class has a higher chance of being right
- Asymmetric cost of error: different cost of wrong prediction



Class Imbalance

- Overall accuracy is most commonly used, but it's insufficient when there's class imbalance because it treats all classes the same
- F1: $2 * (\text{precision}) * (\text{recall}) / (\text{precision} + \text{recall})$



How many retrieved items are relevant?

$$\text{Precision} = \frac{\text{true positives}}{\text{true positives} + \text{false positives}}$$

How many relevant items are retrieved?

$$\text{Recall} = \frac{\text{true positives}}{\text{true positives} + \text{false negatives}}$$

Class Imbalance

- Precision, Recall, and F1?

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Class Imbalance

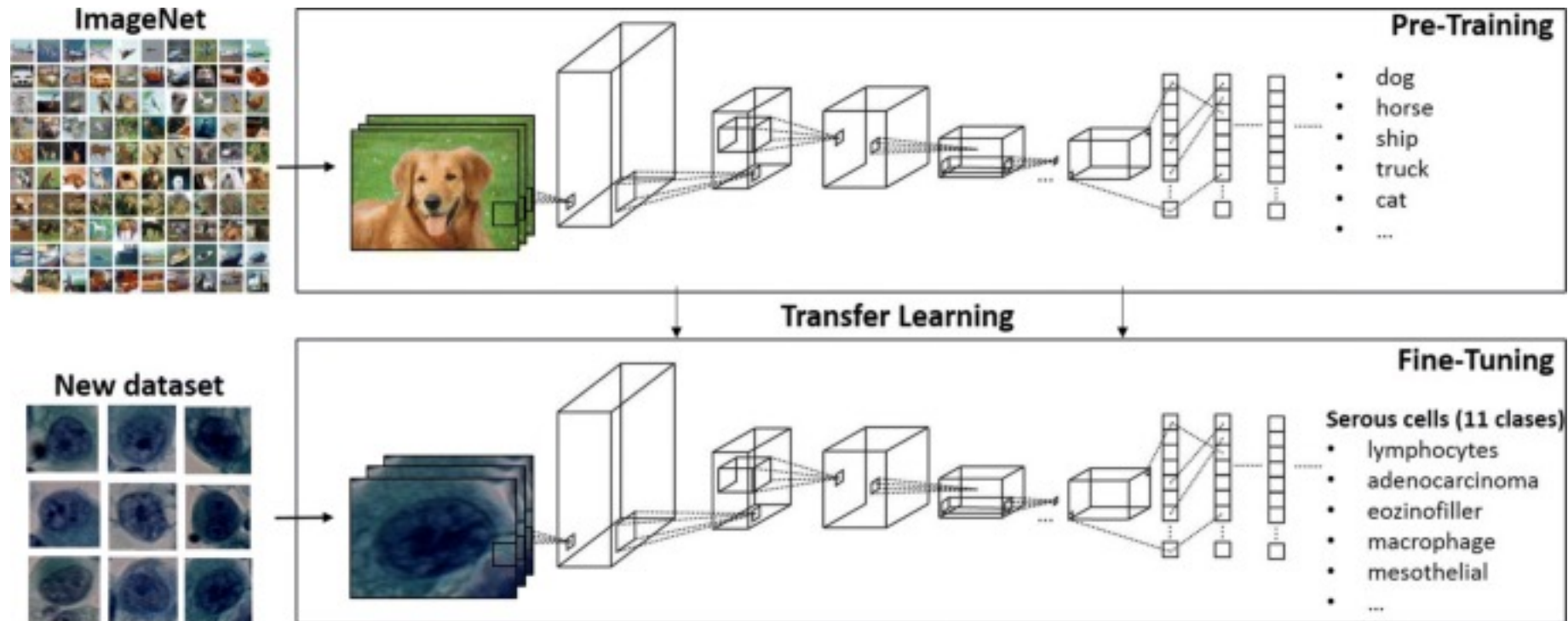
- Model A:
 - Precision = 50% (10/20)
 - Recall = 10% (10/100)
 - F1 = 17%
- Model B:
 - Precision = 50% (90/180)
 - Recall = 90% (90/100)
 - F1 = 64%

Model A	Actual Fraud	Actual Normal
Predicted Fraud	10	10
Predicted Normal	90	890

Model B	Actual Fraud	Actual Normal
Predicted Fraud	90	90
Predicted Normal	10	810

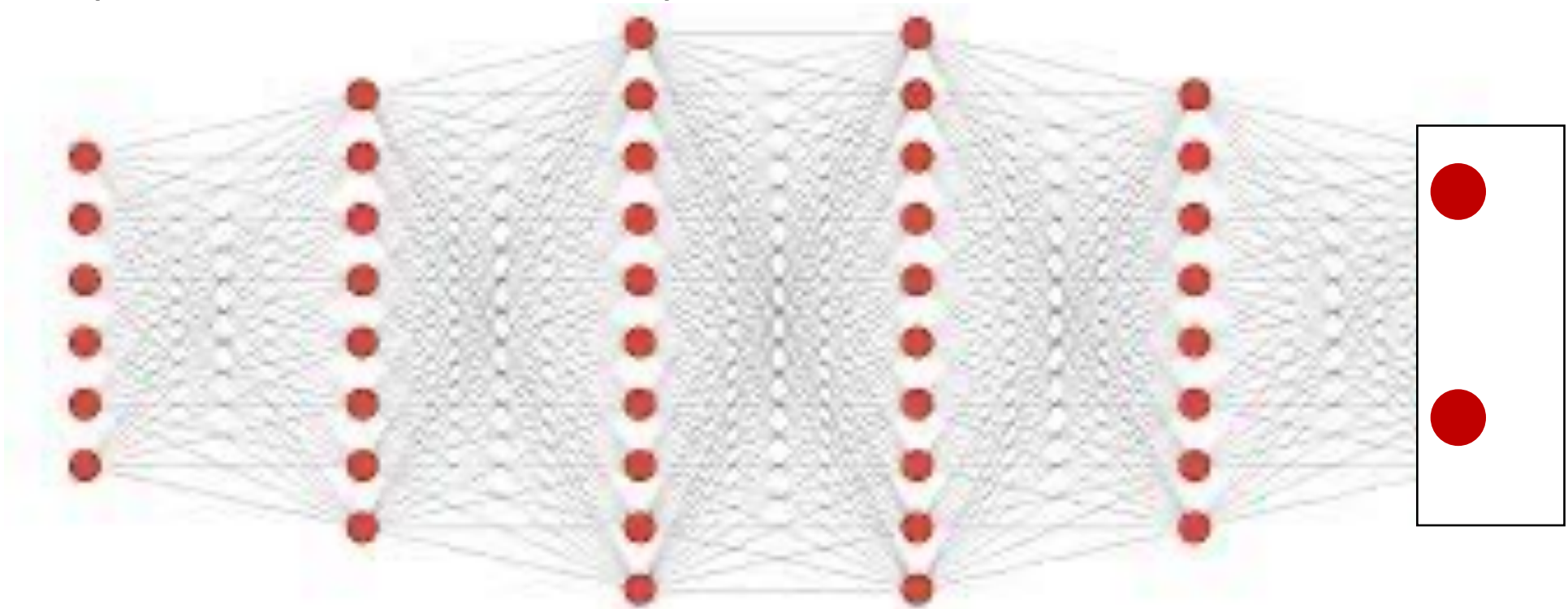
Not Enough Data

- Transfer Learning



To Freeze or Not to Freeze

- While fine-tuning, "freezing" some layers can make training go faster at the expense of lower accuracy



Improving Model Accuracy

- Common culprits:
 - Data
 - **Model**
 - Training Process

Models

- Layers -> generally more will give higher accuracy
- Using a well-established model will generally give higher accuracy – both because structure has been tested and because weights from pre-training are often available
- Details of how model structure (i.e., different layer types, order) affects accuracy is beyond scope of this course

Improving Model Accuracy

- Common culprits:
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 - Model
 - **Training Process**

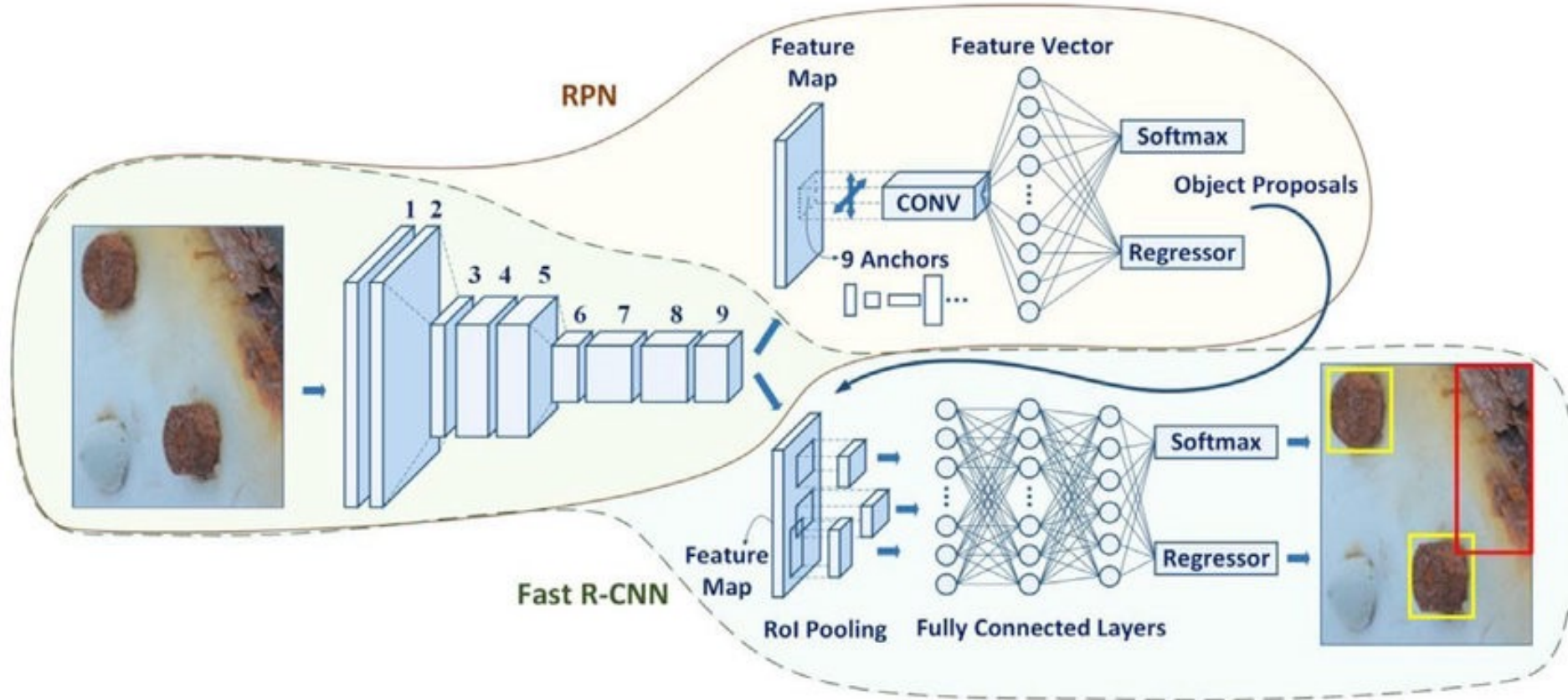
Training Process

- Learning rate
- Optimizer
- Common mistakes: `eval()` vs. `train()`, didn't zero the gradients, didn't pass the correct model's parameters to the optimizer

Object Detection

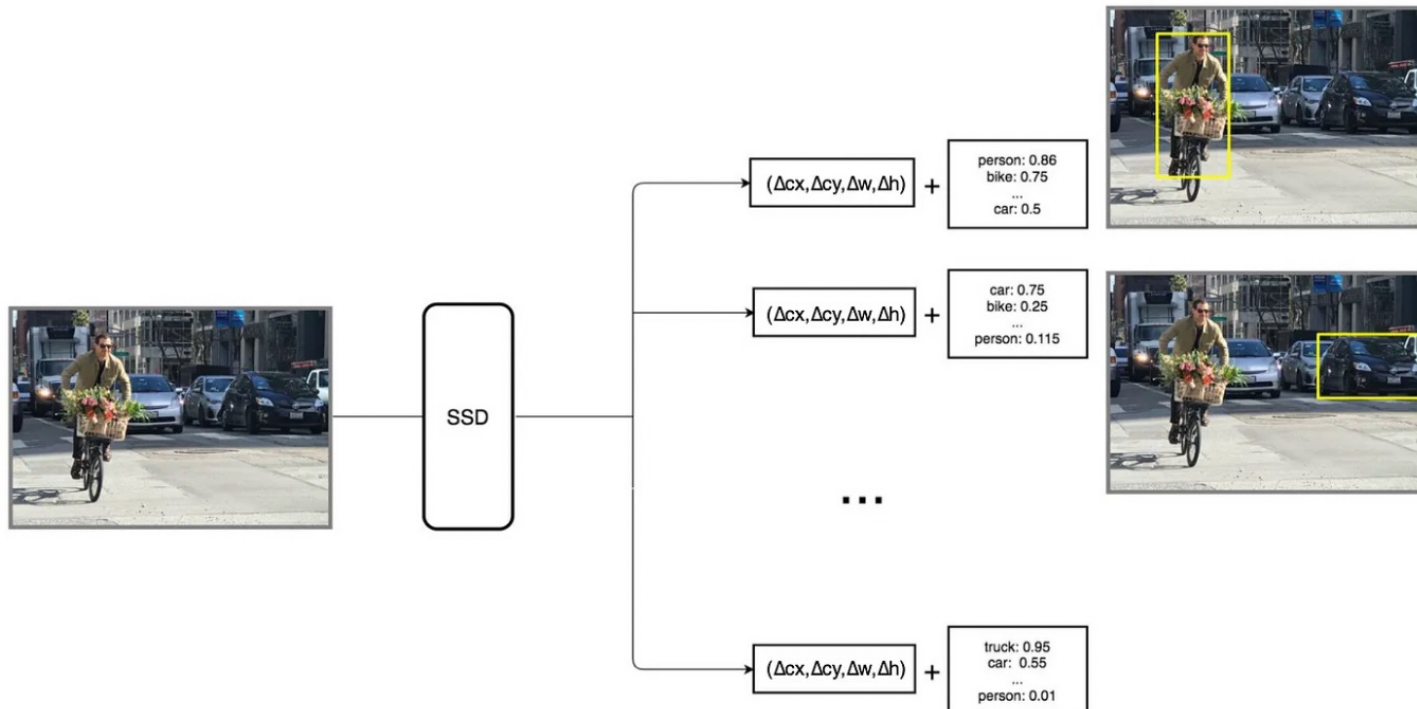
- Need to identify where objects are AND what they are

Faster RCNN

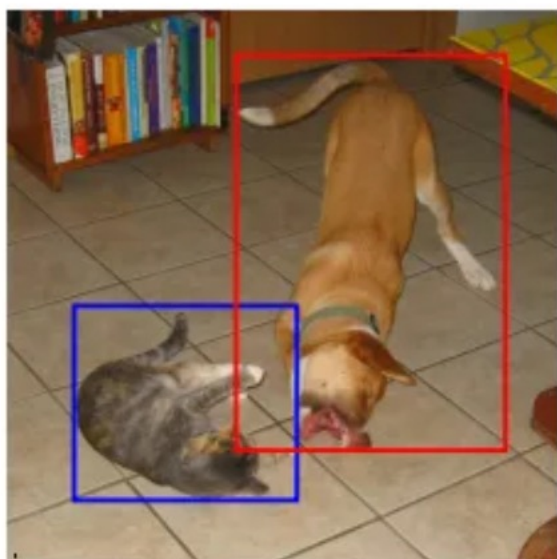


SSD (Single Shot Detector)

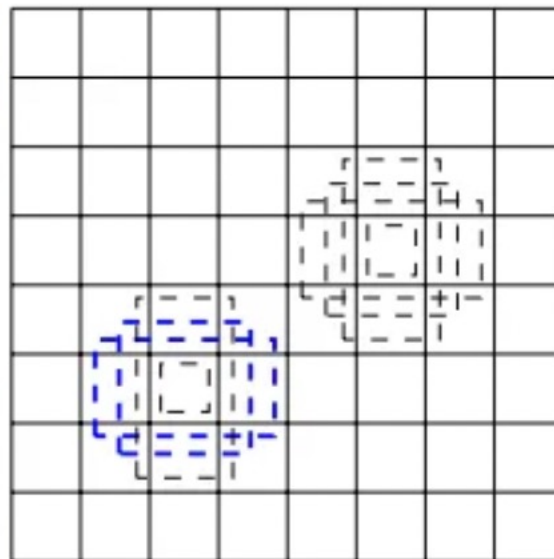
- Alternatively, these could be done together
- Divide image to cells. At each cell, 4 bounding boxes/object pairs are predicted. The object could be None



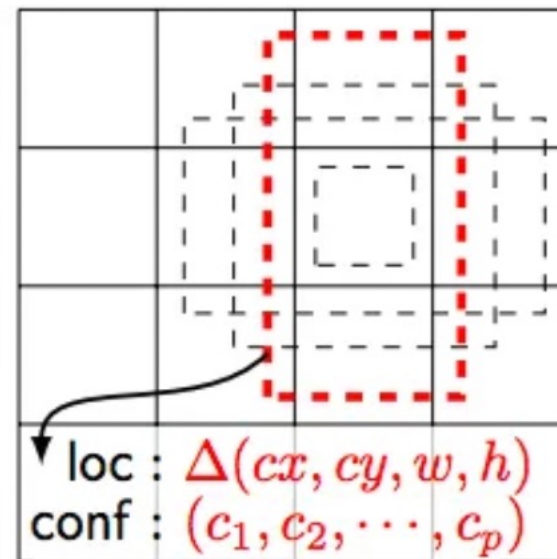
Cells of Different Sizes



(a) Image with GT boxes



(b) 8×8 feature map



loc : $\Delta(cx, cy, w, h)$
conf : (c_1, c_2, \dots, c_p)

(c) 4×4 feature map