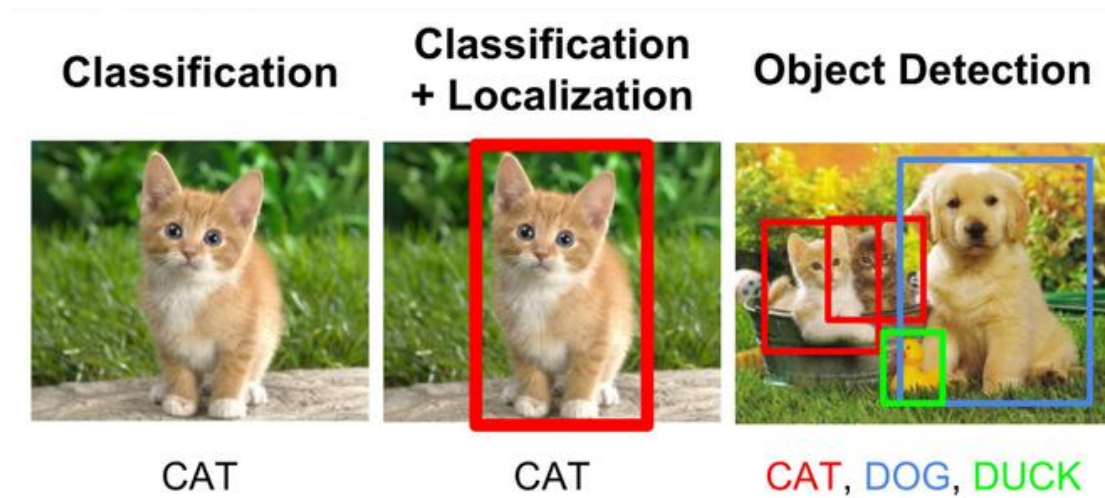


Object localization:



In object Detection there might be multiple objects in the picture and you have to detect them all and localize them

Classification with localization

Classification + Localization: Task

Classification: C classes
Input: Image
Output: Class label
Evaluation metric: Accuracy



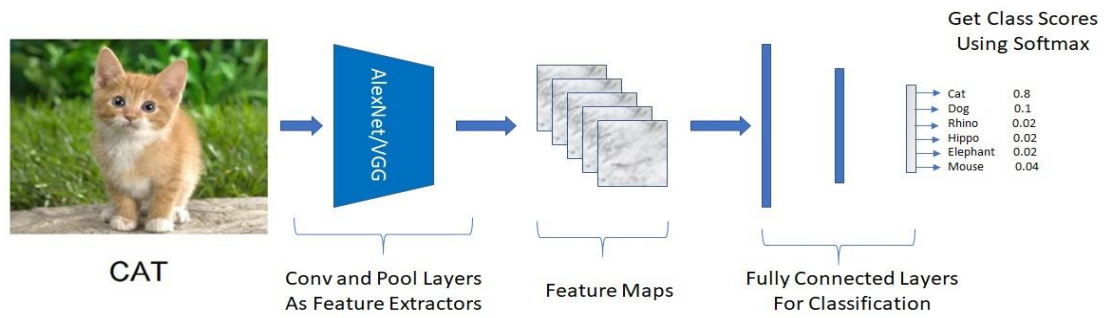
→ CAT

Localization:
Input: Image
Output: Box in the image (x, y, w, h)
Evaluation metric: Intersection over Union



→ (x, y, w, h)

Classification + Localization: Do both



$$Y = \begin{bmatrix} p_1 \\ b_x \\ b_y \\ b_h \\ b_w \\ c_1 \\ c_2 \\ c_3 \end{bmatrix}$$

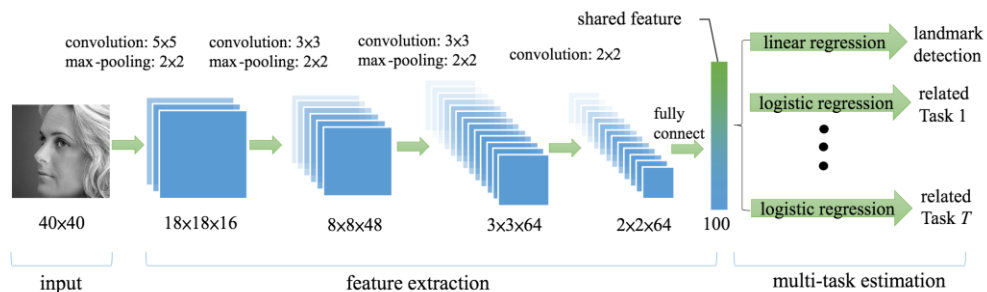
Landmark detection



b_x, b_y, b_h, b_w



$l_{1x}, l_{1y},$
 $l_{2x}, l_{2y},$
 $l_{3x}, l_{3y},$
 $l_{4x}, l_{4y},$
 \vdots
 l_{64x}, l_{64y}

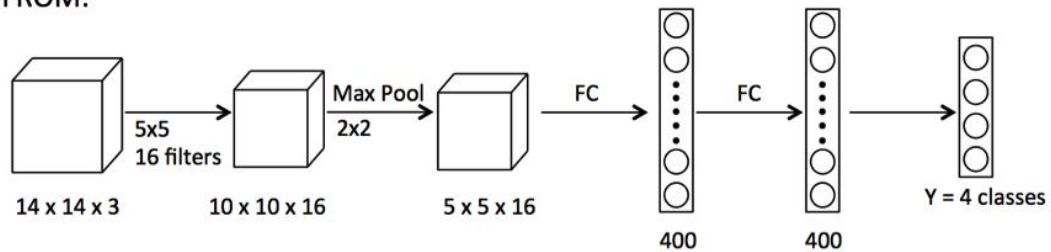


Make object detection by sliding window and change the square size and repeat

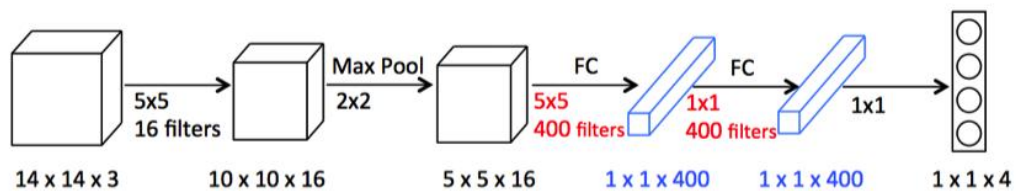
Convolutional implementation of sliding window

Turning fc layers into convolutional layers and change the soft max layer

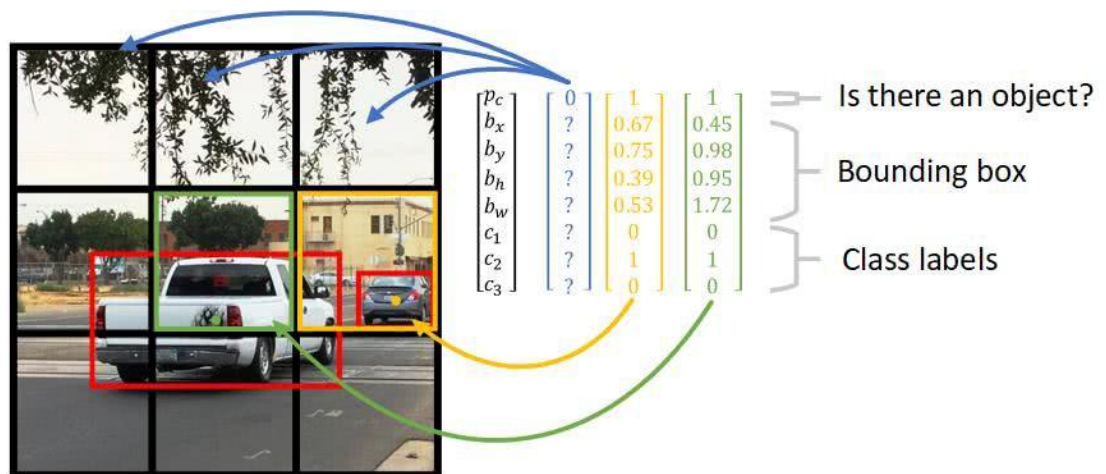
FROM:



TO:



Bounding box prediction




b_x, b_y between 0&1

b_h, b_w could be >1

Intersection over union "IOU"

IOU is a measure of the overlap between two bounding boxes


$$\text{IoU} = \frac{\text{Area of Overlap}}{\text{Area of Union}}$$

Non-max Suppression

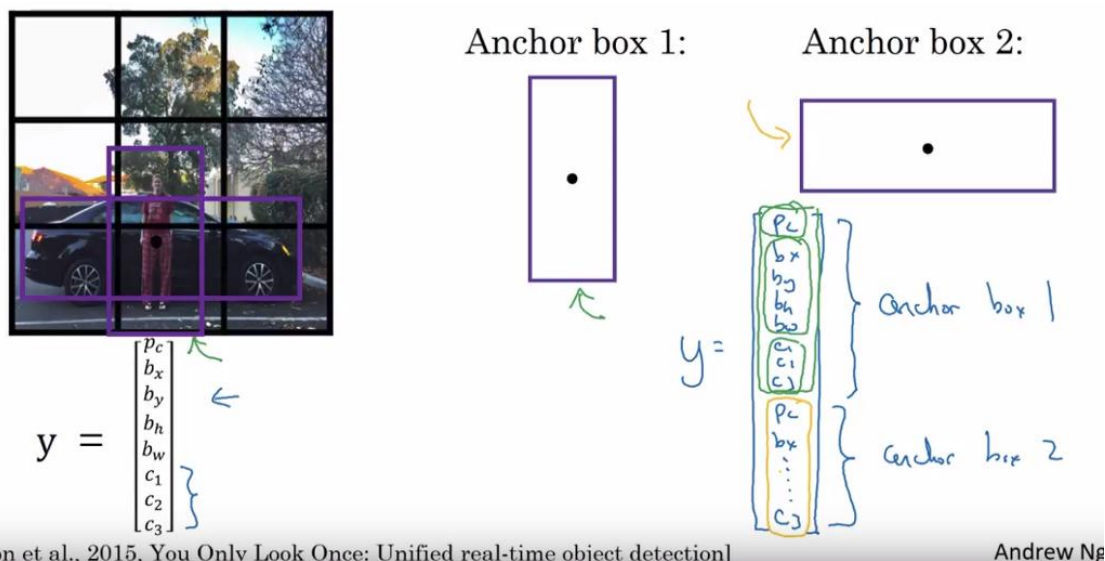
Non-max means that you are going to o/p your maximal probability classification but suppress the close by ones that are non maximal

While the rear any remaining boxes

- Pick box with the largest p_c o/p that as prediction
- Discard any remaining box with $\text{IOU} \geq 0.5$ with box o/p in the previous step

Anchor boxes

One of the problem with object detection is each of the grid cells can detect only one objects so we use the idea of anchor boxes



Week-4 (face recognition)

Face recognition

Demoed both face recognition as well as liveness detection ,the latter meaning making sure that you are a live human

Face verification VS face recognition

Verification

- I/P image ,name ID
- O/P wither the I/P image is that of the claimed person

Recognition

- Has a data base of K persons
- Get an I/P image
- o/p if the image is any of the K person or not recognized

- Face Authentication/Verification (1:1 matching)



- Face Identification/Recognition (1:N matching)



One shot learning:

Learning from one example to recognize the person a gain

Learning a "similarity" function

$d(\text{img1}, \text{img2}) = \text{degree of difference between images}$

If $d(\text{img1}, \text{img2}) \leq \tau$ "same"
 $> \tau$ "different" } Verification



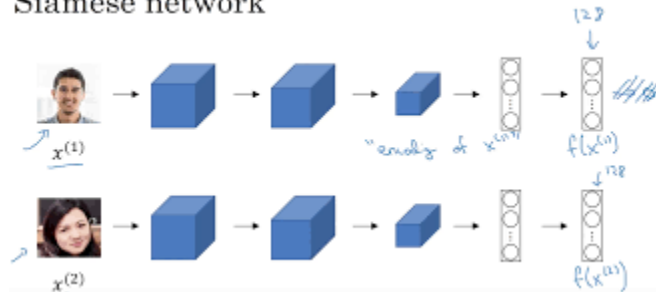
$d(\text{img1}, \text{img2})$

Andrew Ng

Siamese Network

Use to tell us how different is between I/p of two faces

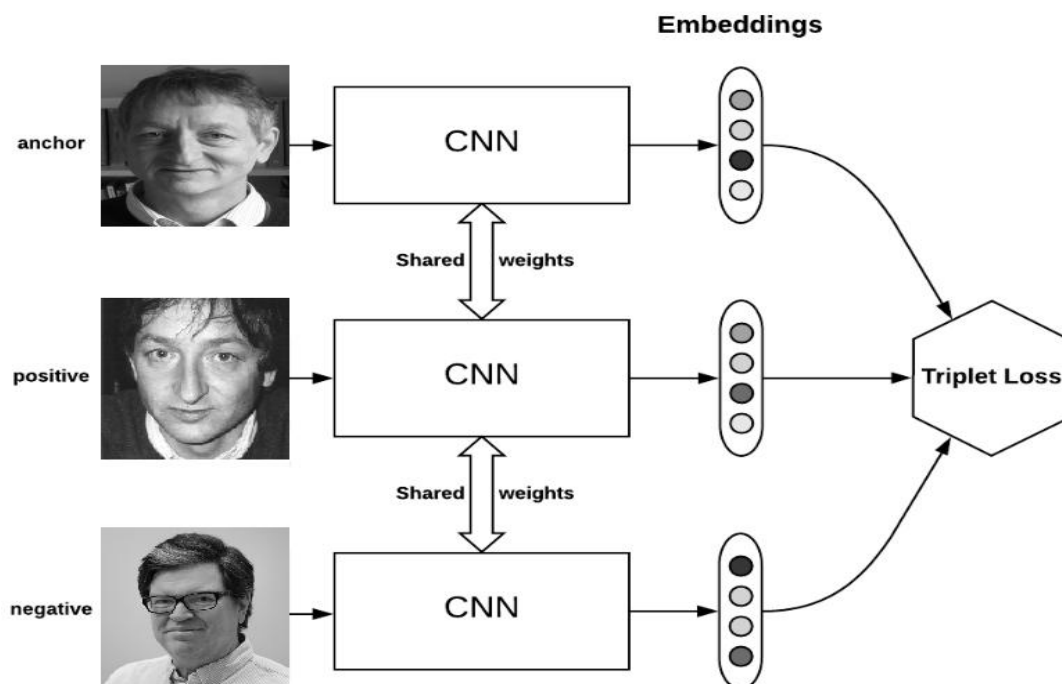
Siamese network

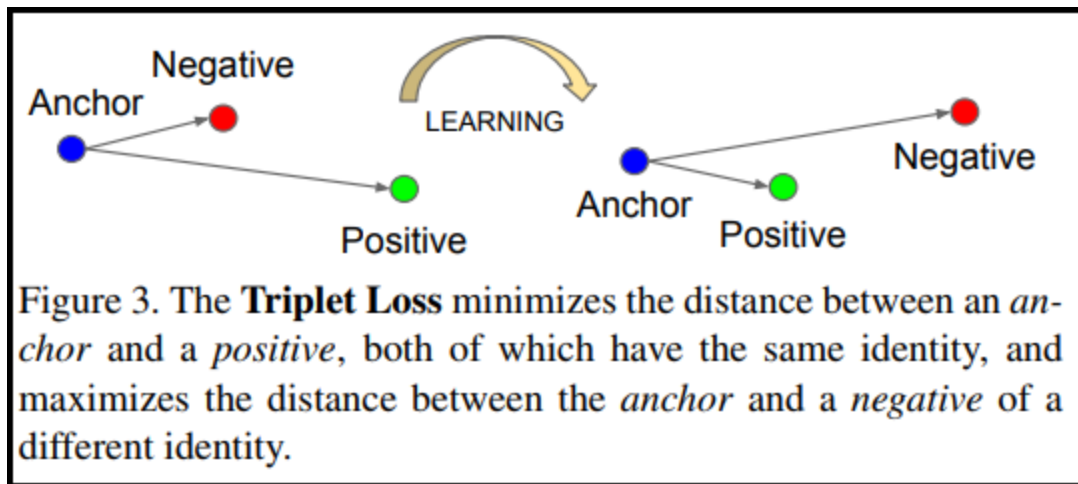


Triplet loss

Learning objective:

You will be looking at anchor image, positive image as well as negative image





We use gradient decent to minimize the cost function J

Face verification and Binary classification

Binary classification gives output as 0 or 1

And the same face verification if both these are the same persons and zero if both of these are for different persons

$$\hat{y} = \sigma \left(\sum_{k=1}^{128} w_k |f(x^{(i)})_k - f(x^{(j)})_k| + b \right)$$