

Solution 1 :

1. Prove that the IoU metric between any two pairs of bounding boxes is always a non-negative real number in $[0,1]$

Let us consider we have two boxes A and B

$$IoU_{metric} = \frac{A \cap B}{A \cup B} \quad (1)$$

From the properties of sets we know that, $A \cup B = A + B - A \cap B$

$$\begin{aligned} \frac{A \cap B}{A \cup B} &= \frac{A + B - A \cap B}{A \cup B} \\ \frac{A \cap B}{A \cup B} &= \frac{A + B}{A \cup B} - 1 \end{aligned} \quad (2)$$

Case 1 : When A and B are disjoint:

$$A \cap B = 0 \text{ (Null)}. \text{ Hence, } IoU = 0 \quad (3)$$

Case 2 : When A and B are slightly overlapping:

$$A + B > A \cup B$$

$$\frac{A + B}{A \cup B} > 1$$

$$\frac{A + B}{A \cup B} - 1 > 0$$

$$\text{Substituting in (2), } \frac{A \cap B}{A \cup B} > 0 \text{ or } IoU > 0 \quad (4)$$

Case 3 : When A \in B or B \in A:

When $A \in B$, $A \cup B = B$ and $A \cap B = A$

$$\text{Since } A < B, \frac{A \cap B}{A \cup B} < 1 \quad (5)$$

Similarly when, $B \in A$, $A \cup B = A$ and $A \cap B = B$

$$\text{Since } B < A, \frac{A \cap B}{A \cup B} < 1 \quad (6)$$

Case 4 : When A and B are totally overlapping:

In this case, if $A = B$, $A \cup B = A$ and $A \cap B = A$

$$\frac{A \cap B}{A \cup B} = 1 \quad (7)$$

Combining, (3), (4), (5), (6), (7),

$$0 \leq \text{IoU metric} \leq 1 \quad (8)$$

Hence, proved.

2. If we represent each bounding box as a function of the top-left and bottom-right coordinates (assume all coordinates are real numbers) then argue that the IoU metric is non-differentiable and hence cannot be directly optimized by gradient descent.

In order for a function to be differentiable, it must be continuous, and its derivative must be continuous too. Considering a case for 2 bounding boxes, where B is fixed and A is variable, we can have 4 different possibilities as mentioned in Part 1 of the Solution 1.

The IoU varies significantly and is not continuous.

Case 1: When A and B are disjoint - $\text{IoU} = 0$

Case 2: When A and B are slightly overlapping - $\text{IoU} > 0$

Case 3: When $A \in B$ or $B \in A$ - $\text{IoU} < 1$

Case 4: When A and B are totally overlapping - $\text{IoU} = 1$

Therefore, we can see that the function is not continuous (smooth) and hence is not differentiable and hence cannot be directly optimized by gradient descent.

Solution 2 :

1. Calculate A, B, C, D, E, F

A = 30 - Since padding = same, image size remains constant

B = 30 - Since padding = same, image size remains constant

C = 16 - Same as number of filters applied

D = 30 - Dimensions must be similar to regular conv layer and also as padding = same, image size remains constant

E = 30 - Dimensions must be similar to regular conv layer and also as padding = same, image size remains constant

F = 64 - Same as number of filters applied

2. Computational cost of the regular conv layer -

$$(30 \times 30 \times 64) \times (5 \times 5 \times 200) = 288,000,000$$

Computational cost of bottleneck path -

$$(30 \times 30 \times 16) \times (1 \times 1 \times 200) + (30 \times 30 \times 64) \times (5 \times 5 \times 16) = 25,920,000$$

$$\text{Ratio} = 25,920,000/288,000,000 = 9/100 = 0.09 = 9\%$$

Hence, using by using bottleneck, we reduce the computations to 9% of the regular conv layer

Solution 3 :

In attached ipynb file.

Observations with COCO Keypoint Person Detector model with a ResNet50 – FPN base and following thresholds –

Threshold = 0.1 - The number of silhouettes observed were maximum, but it also misclassified other items as persons (like a couple of cotton candies on top). This can be reduced by increasing the threshold we can avoid these misclassifications.

Threshold = 0.2 - The observed misclassification had a confidence of 15% and hence by increasing the threshold to 20% we were able to omit such cases (for the provided picture).

Threshold = 0.5 - The number of silhouettes were same as before and the misclassifications were avoided as the threshold is high.

Threshold = 0.7 - The number of silhouettes were lower than before.

Threshold = 0.9 - The number of silhouettes were even lower than before.

Hence, I chose the threshold of 0.5 for the final output

Observations with Mask R – CNN model with ResNet50 – FPN and following thresholds

Threshold = 0.1 - The number of misclassifications were quite high like the cotton candies misclassified as umbrella and lamp as traffic light.

Threshold = 0.3 - The number of misclassifications were lower than before but still quite high like the cotton candies misclassified as umbrella.

Threshold = 0.5 - The number of misclassifications were still quite high but lost on correct classification like cycle

Threshold = 0.55 - The number of misclassifications were lower.

Threshold = 0.9 - The number of misclassifications were even lower than before but missed out on some correct classifications.

Hence, I chose the threshold of 0.55 for the final output although misclassification of balloons could not be avoided.

After training the balloons –

At threshold of 0.5, most of the balloons were classified and classified correctly.

References -

<https://github.com/facebookresearch/detectron2>
<https://gilberttanner.com/blog/detectron-2-object-detection-with-pytorch>
https://colab.research.google.com/drive/16jcaJoc6bCFAQ96jDe2HwtXj7BMD_-m5#scrollTo=ZyAvNCJMmvFF
