

Q1.  $\neq$  mAP is evaluation metric for object detection model. It measures how well a model performs in detecting objects across different categories by considering both precision and recall.

$$\text{Precision} = \frac{TP}{TP + FP}$$

where, TP = True Positive

FP = False Positive

$$\text{Recall} = \frac{TP}{TP + FN}$$

where, FN = False Negative

~~Avg~~ Average Precision (AP):-

Avg Precision is computed from the precision-recall curve. The curve is plotted with recall values on the x-axis & precision values on the y-axis. AP is the area under this curve.

Mean Average Precision (mAP):-

mAP is a mean of AP values for all classes. If there are  $N$  classes, it is calculated

$$mAP = \frac{1}{N} \sum_{i=1}^N AP_i$$



## ~~RCNN~~ Fast RCNN

Q2.	Features	RCNN	Fast RCNN	Fast RCNN
①	Region Proposals	Selective Search	Selective Search	Region Proposal Network (RPN)
②	Feature Extraction	For each Proposal	Single Feature map	Single Feature map
③	Classification & Regression	Sum & linear regression	Integrated into network	Integrated into network
④	Test on particular image	46 sec	2 sec	2 milliseconds

Q3. The RPN in Fast RCNN significantly improves the efficiency of the detection process through several key optimizations.

- ① In Fast RCNN the RPN shares convolutional layers with the main detection network.
- ② For each position in the Feature map,



RPN generates multiple anchor boxes of different scales & aspect ratios leading to 9 anchors per position.

- ③ For each anchor, the RPN simultaneously predicts an objectness score & defines the anchor's coordinates through bounding box regression.
- ④ Optimizing the entire network for both region proposal & detection tasks through unified training.
- ⑤ Reducing the no. of proposals via Non-Maximum Suppression, minimizing computational load in detection stage.

Q9.  $\neq$  IOU measures the overlap b/w two bounding boxes: a predicted bounding box and a ground truth bounding box. It is defined as the ratio of the area of the intersection of the two bounding boxes to the area of their union.

$$IOU = \frac{\text{Area of Intersection}}{\text{Area of Union}}$$

Importance of IOU :-

- ① It helps in setting thresholds for evaluating ~~the~~ true positive & false positive.



- ⑪ It is used in loss functions during training to improve model performance.
- ⑫ It plays a key role in post-processing steps like Non-Maximum Suppression.

Q9. In object detection, multiple bounding boxes can be often be proposed for the same object. These bounding boxes usually have different confidence scores indicating the containing an object. NMS helps in selecting the most accurate bounding box for each object. ~~Following~~

Significance of NMS:-

- ① ~~Redundancy~~ Redundancy Reduction
- ② Improved Accuracy
- ③ Efficiency