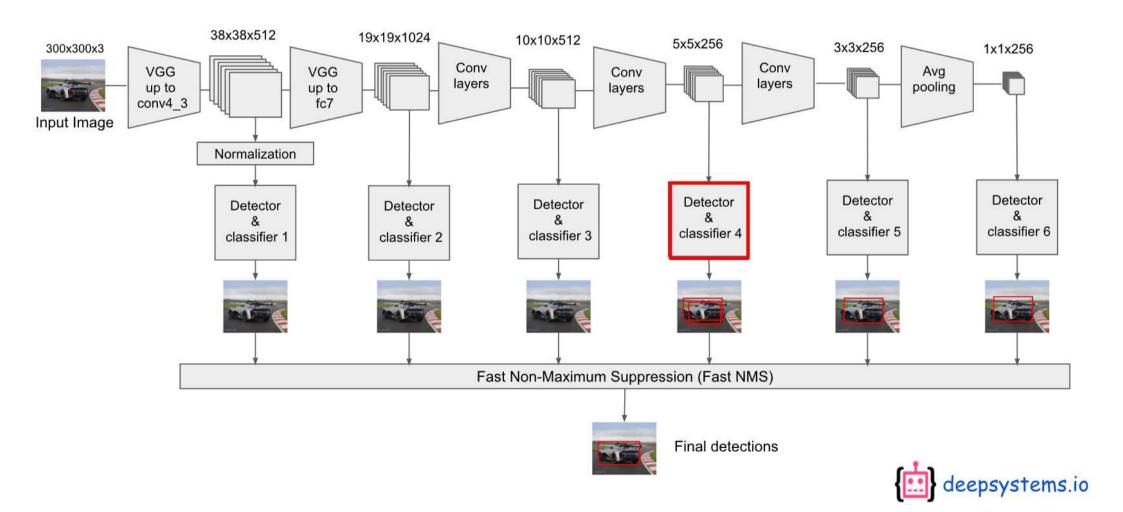
# SSD Single Shot MultiBox Detector

박 철

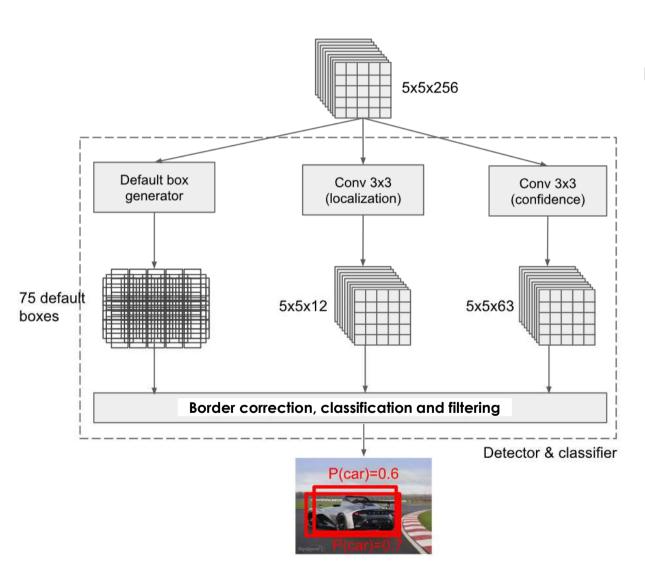
### SSD 300 architecture

base Network : VGG-16 conv4\_3까지 적용

multi feature maps: 38x38, 19x19, 10x10, 5x5, 3x3, 1x1 피쳐맵



### SSD 300 Detector & classifier architecture



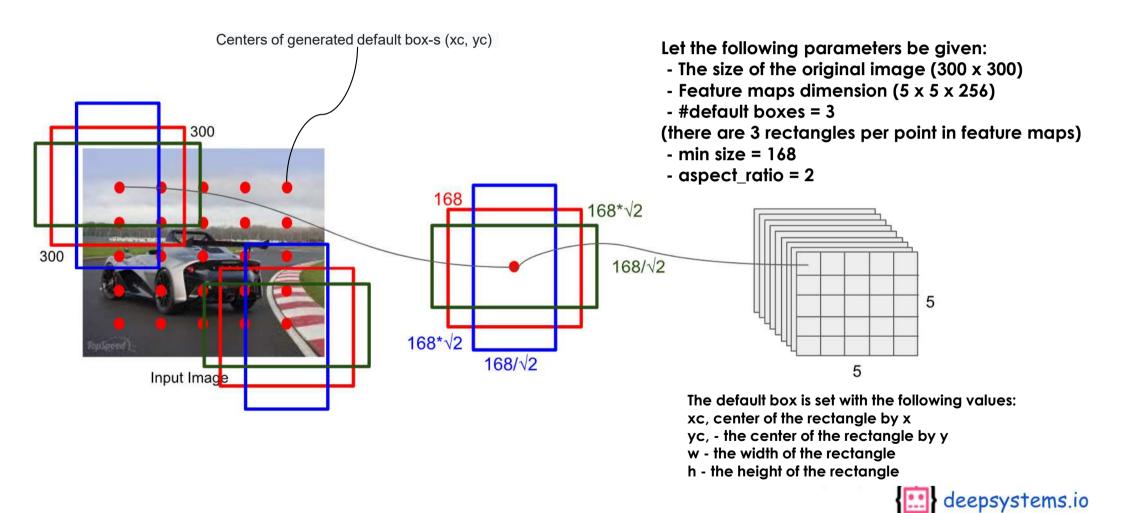
Let the following parameters be given:

- The size of the original image  $(300 \times 300)$
- Feature maps dimension (5 x 5 x 256)
- #default boxes = 3

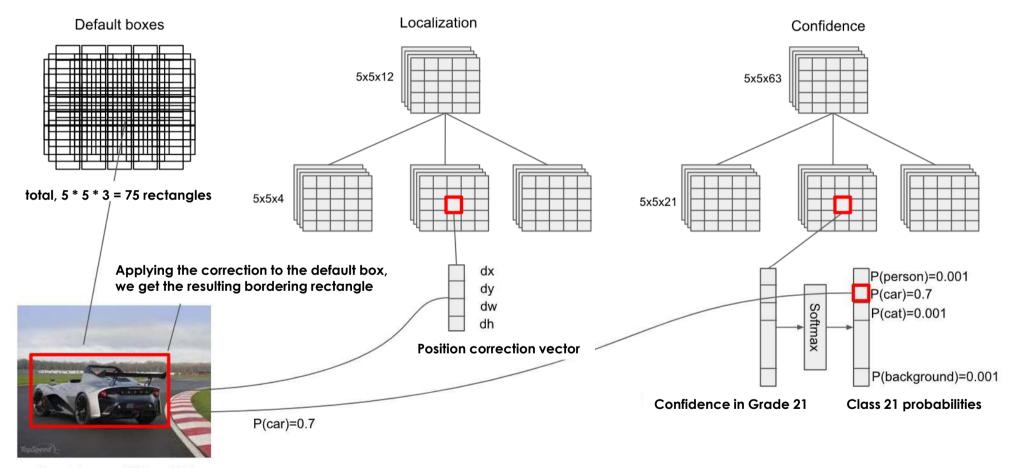


## Generating default boxes

In total, 5 \* 5 \* 3 = 75 rectangles will be generated



## Border correction, classification and filtering 1

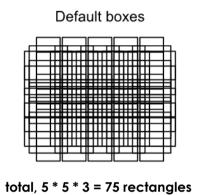


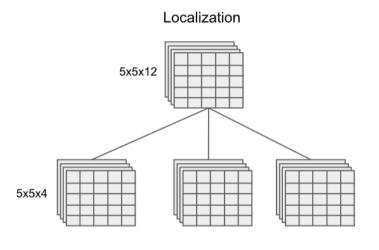
Input Image (300 x 300)

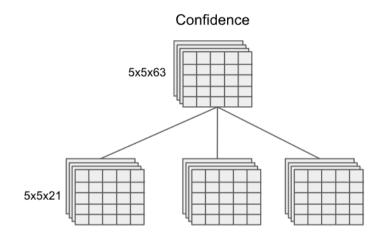
Result: 75 children bound to the class

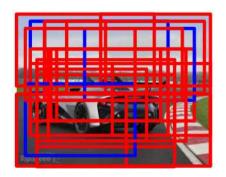


## Border correction, classification and filtering 2









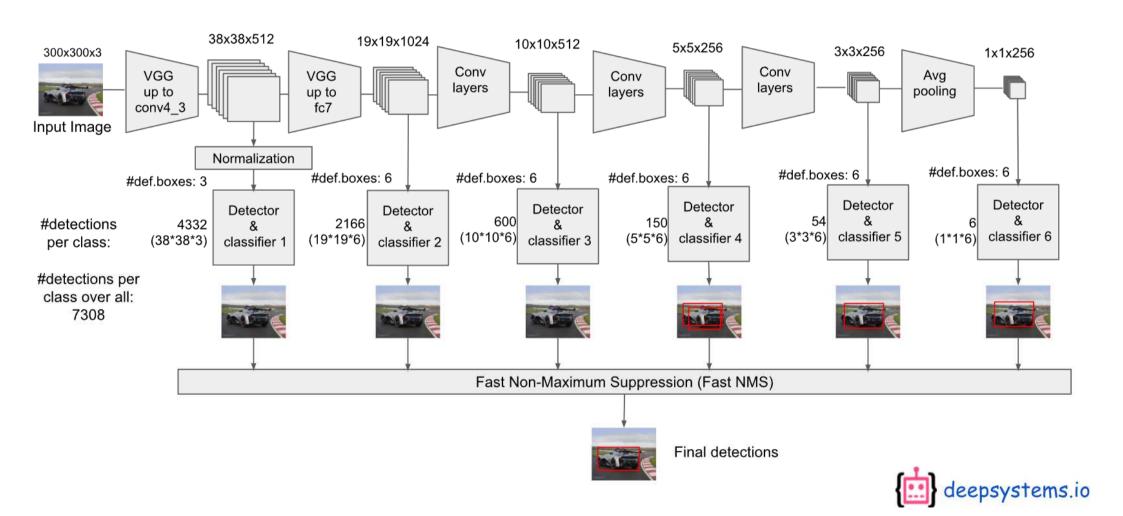
75 detectable objects

Filtering by confidence



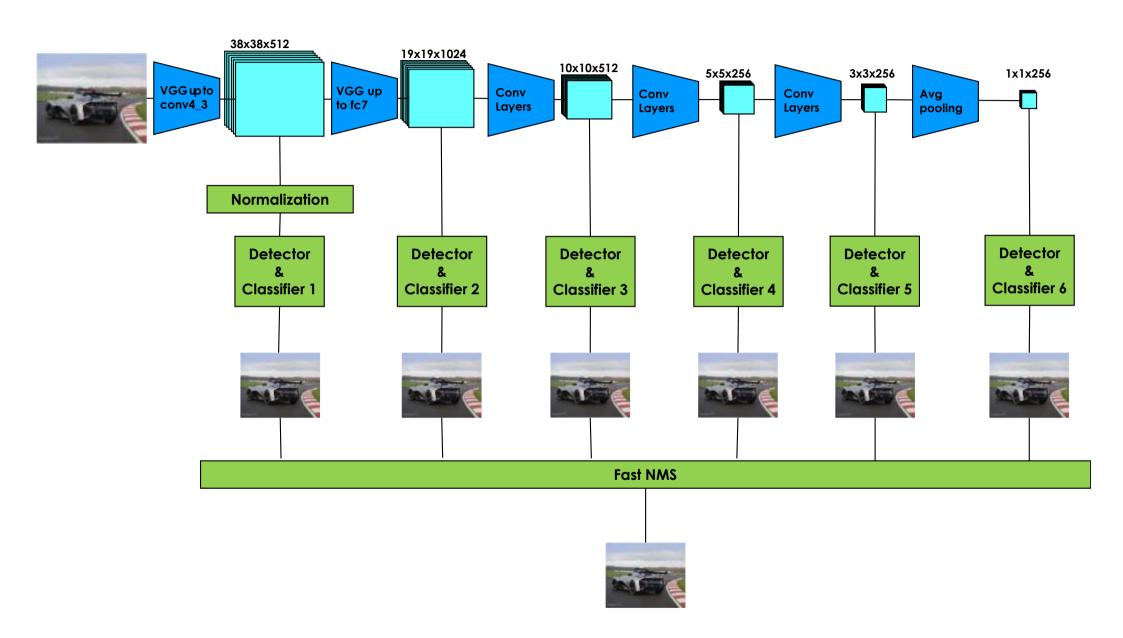


### SSD 300 architecture



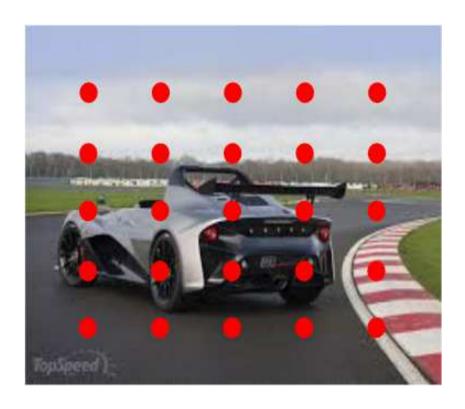
### **Key Points**

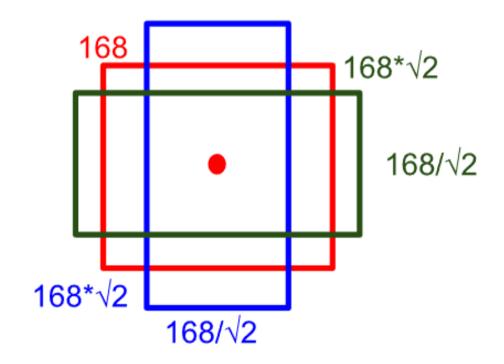
- SSD architecture enables real-time object detection
- The quality of work is close to Faster R-CNN
- Detection occurs at different scales, which allows you to localize objects of different sizes
- A large number of default boxes are used, covering the input image at different scales
- At the Inference stage, the SSD 300 architecture detects 7308 objects, most of which are subsequently filtered.

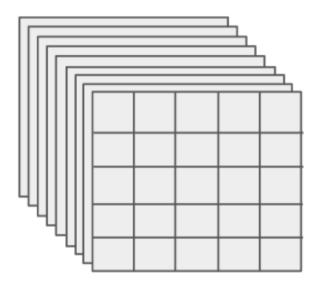


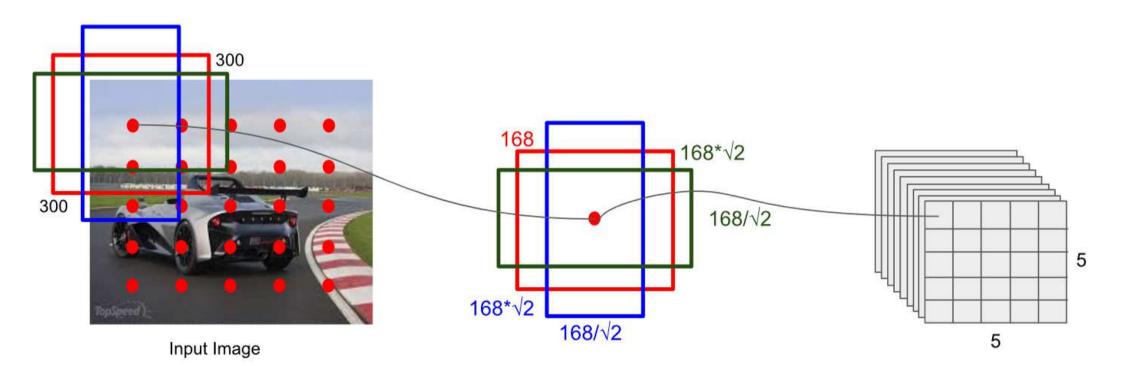












#### 과정 개요

아웃 풋을 만드는 공간을 나눈다(multi feature map). 각 피쳐맵(아웃풋맵)에서 다른 비율과 스케일로 default box를 생성하고 모델을 통해 계산된 좌표와 클래스값에 default box를 활용해 최종 bounding box를 생성한다.

정확도 향상을 위해 서로 다른 피쳐맵에서 서로 다른 스케일의 예측을 할 수 있게 했다. (Yolo v1은 최종 아웃풋은 한 피쳐맵이고, 각 그리드 셀당 2개의 바운딩 박스를 예측하는데 비해 SSD는 여러가지의 그리드셀(피쳐맵)을 가지고 각 피쳐맵당 여러가지의 (보통 6개)바운딩 박스를 가지기 때문) 이 역시 2.1 Model과, 2.2 Training에서 자세히 살펴볼 것.