

Object

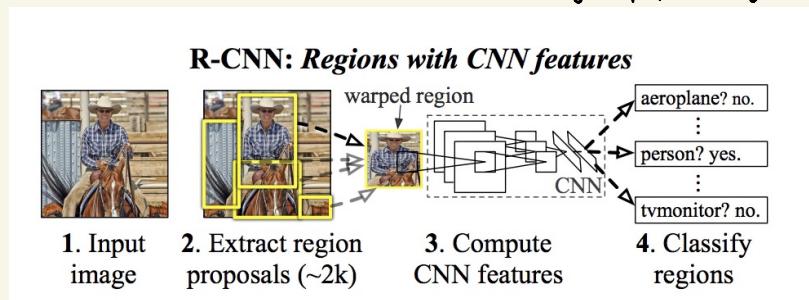
Detection

Object detection algorithm

R-CNN:

extract 2000 regions from raw image.

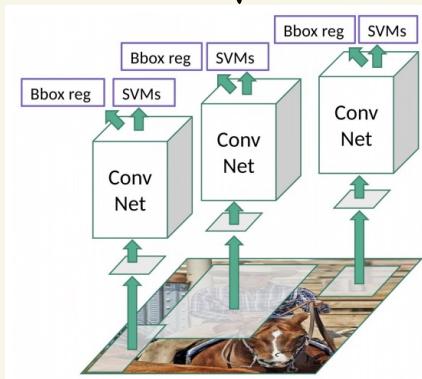
region proposals (2000 region proposals generated by selective search algorithm)



↳ 2000 candidate region proposals warped into a square, fed into a CNN.
CNN produces a 4096-dim feature vector as output.

Features feed into SVM to classify

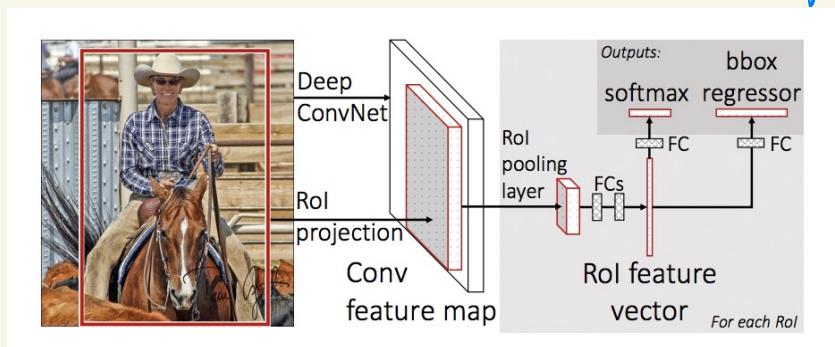
Predict 4 offset values of bounding box.



problems with R-CNN:

- ① waste of time to train (classify 2000 region per img)
- ② Not suitable for Real-time (around 47s per img)
- ③ Selective search algorithm is fixed. (could generate bad region proposals)

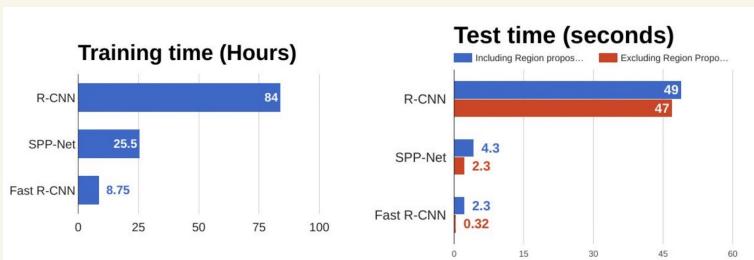
Fast R-CNN → solve drawbacks of R-CNN.



(R-CNN) feed region proposals to CNN

(Fast R-CNN) feed input img to CNN → generate convolutional feature map

↓
{ identify the region of proposals
wrap them into squares } → reshape into fixed size (RoI pooling layer) → feed into FC layer → { softmax
bbox reg }



No need to feed 2000 region proposals to CNN

Convolution operation Once per img.

region proposals : bottlenecks in Fast R-CNN

Faster R-CNN

slow, time-consuming

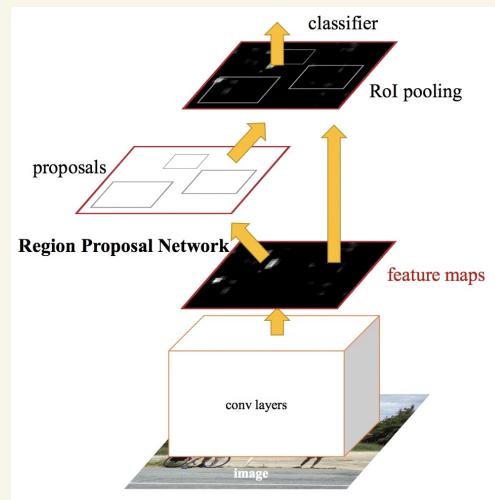
R-CNN & Fast R-CNN: Selective search

Faster R-CNN : "eliminated"

img $\xrightarrow{\text{CNN}}$ Convolutional feature map

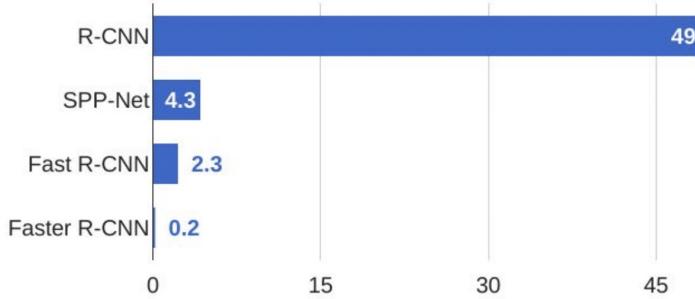
~~Selective search~~

\downarrow predict the region proposals $\xrightarrow{\text{ROI pooling layer}}$



reshape to fixed size \longrightarrow { classify
predict offset values}

R-CNN Test-Time Speed



Much faster than before

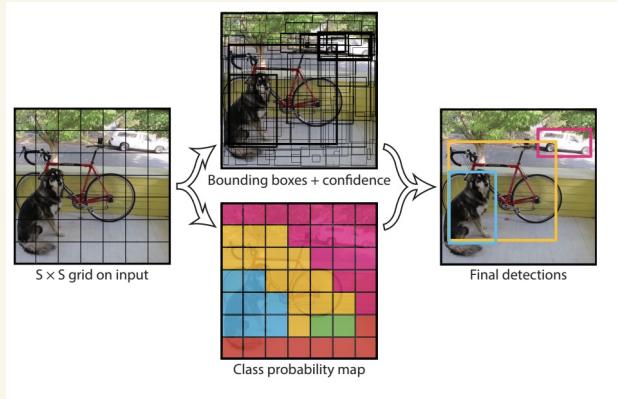
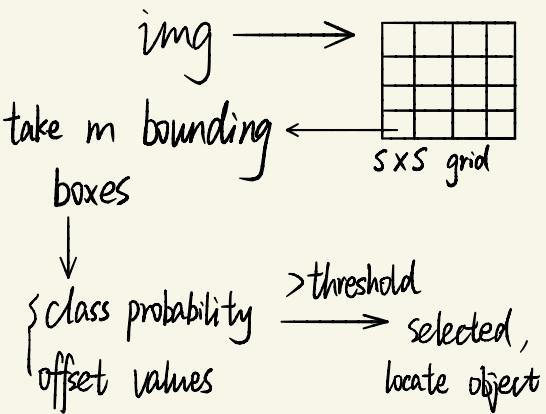
suitable for real-time object-detection

YOLO

YOLO — You Only Look Once

R-CNN, Fast R-CNN, Faster R-CNN : use regions to localize object.

YOLO : a single CNN predicts { bounding boxes }
class probabilities for boxes



far more faster (45 fps)

struggle with small objects.