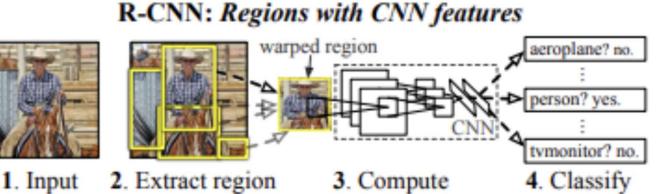
리 R-CINI®

Rich feature hierarchies for accurate OD and SS

허지혜발표



CNN features

- 1. takes an input image
- 2. look around 2000 bottom-up region proposals

image

3. computes feature for each proposal using a large CNN

proposals (~2k)

4. classifies each region using SVMs

mAP 53.7% on PASCAL VOC 2010

regions

In other word, Our object detection system consists of three models.

1) Generates category-independent region proposals. (독립적인 범주의 지역 제안 생성)

These proposals define the set of candidate detections available to our detector

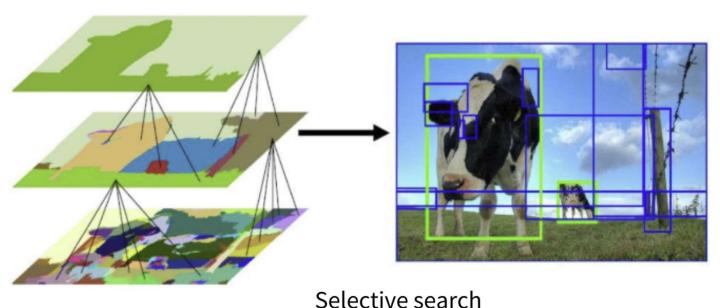
- 2) Large convolutional neural network that extracts a fixed-length feature vector from out region.
- 3) A set of class-specific linear SVMs.

Test: PASCAL VOC 2010-12

Detail gow their parameters are learned and show resulus on test dataset.

Region proposals.

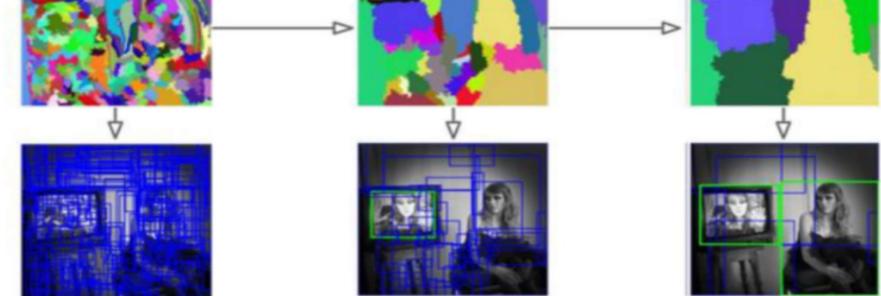
We use **selective search** to enable a controlled comparison(비교) with prior detection work.



- 1) color, size et al => small segmented areas
- 2) Bottom-up 방식으로 small s egmentatin areas를 합쳐 big s egmented areas를 만든다.
- 3) 위 작업을 반복해 2000개의 reg ion proposal를 생성한다. (ROI(Regional of Interest)를 기 준으로 생성)

Bottom-up segmentation, merging regions at multiple scales

Convert regions to boxes



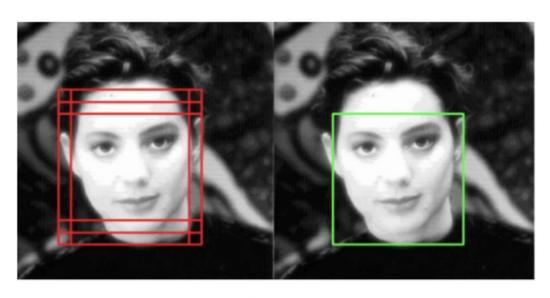
Selective search

- < 1. Supervised Pre-Training >
- CNN의 pre-train에 사용한 데이터 : Imagenet
- caffe 코드로 pre-training 수행
- < 2. Domain-Specific Fine-Tuning >
- CNN을 새로운 detection과 wraped proposal window에 adapt하기 위해, wraped proposals를 사용해 CNN 파라미터 학습에 SGD 사용
- CNN 아키텍쳐는 object의 class가 변경됨에 따라 마지막 레이어가 수정된 것 외에는 동일
- IoU 임계값은 0.5사용
- 학습률 0.001

Feature extraction

Make 2000 region proposal -> resize (227,227) RGB image (wrapping all pixel 비율 고려 x)

- => Through five convolutional layers and two fully connected layers for training Imagenet
- => Gain 4096-dimensional feature vextor



[도표 3] Non Maximum Supperssion

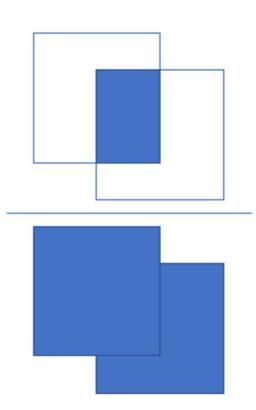
SVM을 통과하여 각각의 박스들은 어떤 물제일 확률값을 가질 수 있다.

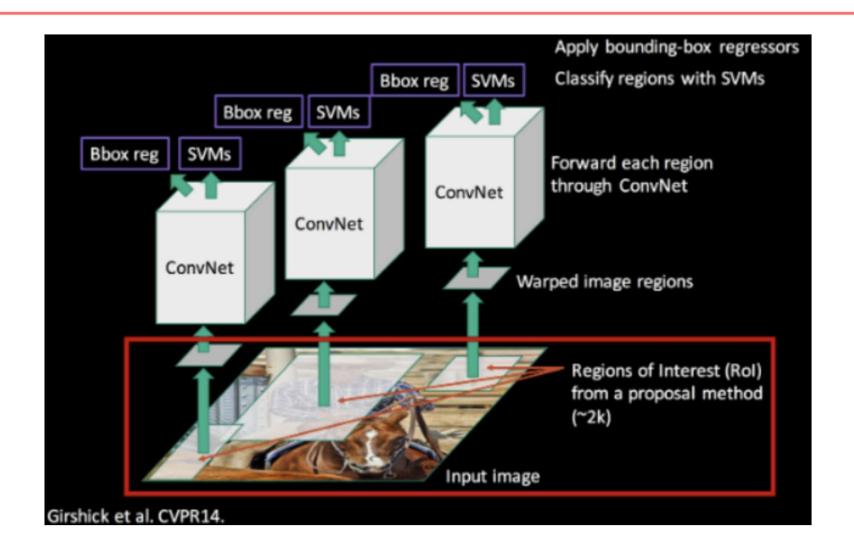
그 확률값 중에 가장 큰 값을 제외하고 나머지 박 스들을 삭제해야 한다.

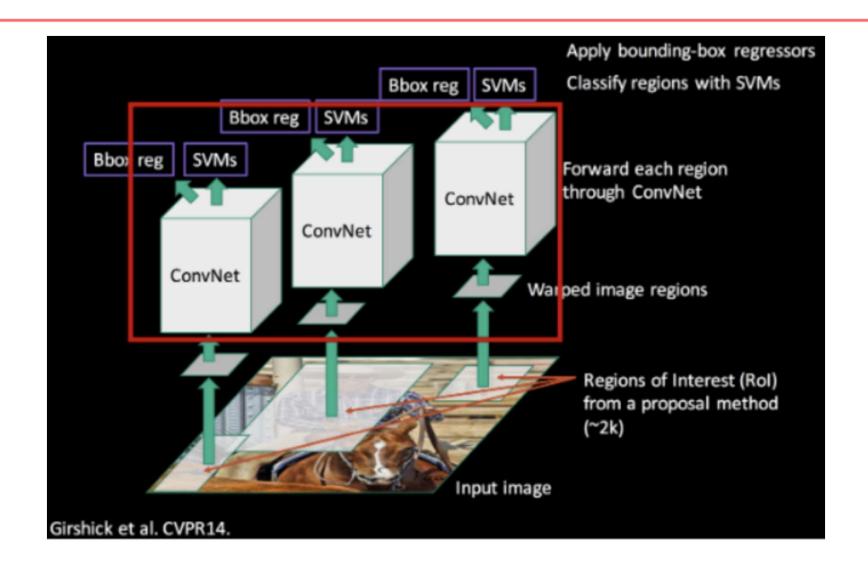
이때 각 박스가 동일한 물체를 나타내는지 판별하기 위해 IoU 값을 0.5보다 큰 값을 뽑아낸다.

IoU(Intersection over Union)

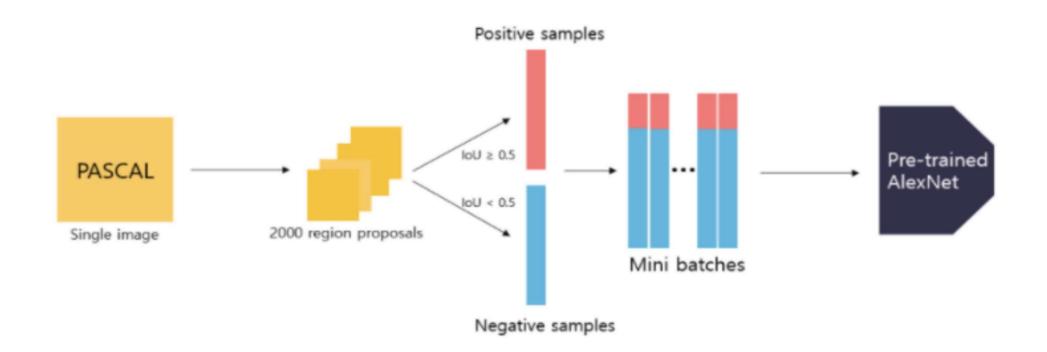
= Intersection / A + B - Intersection

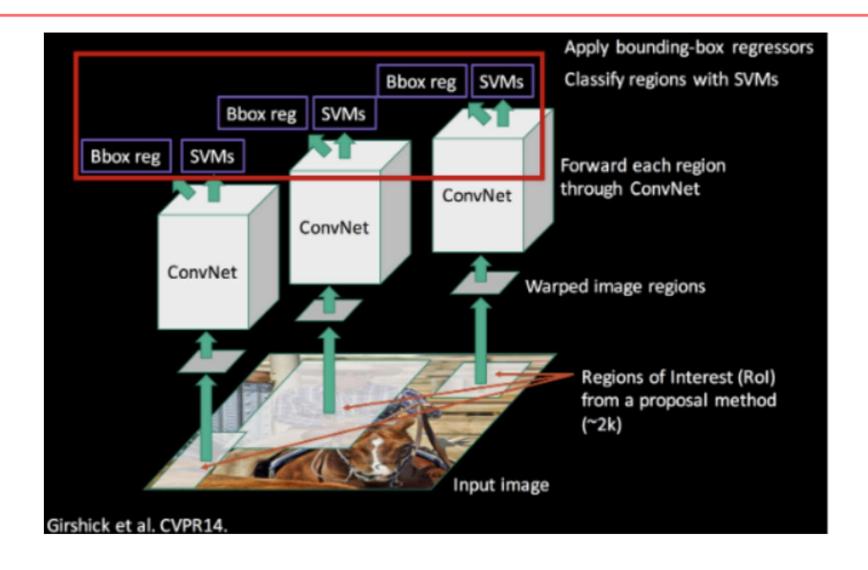


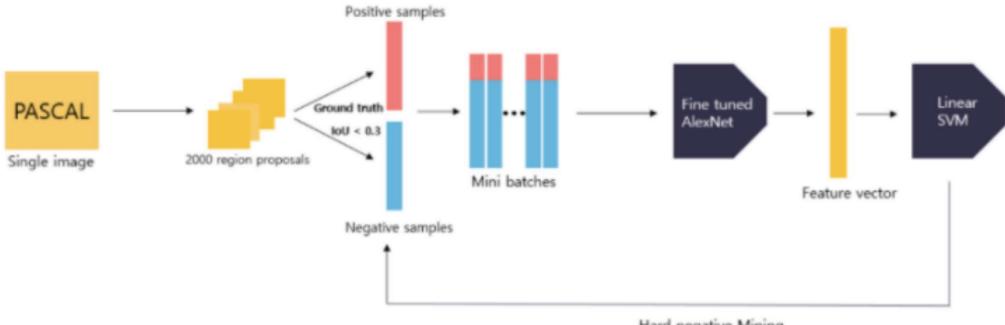




Domain-specific fine-tuning







Hard negative Mining

Training

감사합니다®

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