

- **#paper/to-read** ~ 2013 CE ~ Object Detection, Semantic Segmentation
 - **Rich feature hierarchies for accurate object detection and semantic segmentation**
 - <https://arxiv.org/abs/1311.2524>
 - Sequel papers:
 - [Fast R-CNN](#)
 - [Faster R-CNN](#)
 - [Mask R-CNN](#)
 - [Context R-CNN](#)
 - Mentioned papers:
 - [Regionlets](#)
 - [Objectness of Image Windows](#)
 - [Recognition Using Regions](#)
 - [Category Independent Object Proposals](#)
 - [DeCAF](#)
 - [Adaptive Deconvolutional Networks](#)
 - [Diagnosing Error in Object Detectors](#)
 - Mentioned topics:
 - [Deformable Part Models](#)
 - [Hard Negative Mining](#)

• Summary

- **Architecture:** input image → extracting region proposals (~2k) → feature-extraction [CNN](#) → class prediction or bounding box regression [SVM](#) (or both).
 - [Intersection Over Union, IoU](#) with ground truth labels should be greater than 0.5.
 - At test time, [IoU](#) is used to select the dominating prediction for a region.
 - With the enabled bounding box [Regression](#), the model learns to predict shifts of the predicted bounding box relative to the coordinates of the proposed region.

- R-CNN can scale to thousands of classes without resorting to hierarchical tricks or [Hashing](#).
 - **Training process** consists of two parts.
 - Supervised pre-training happens on a big [Classification](#) dataset to make the CNN part learn without overfitting on (usually) small detection datasets.
 - Domain-specific fine-tuning is used to train SVM heads on region proposals.
 - Those proposals should fit 227×227 pixels so they are usually warped.
 - Different CNN architectures yield different [mAPs](#) but also require different amount of [Compute](#).
 - For region proposals, [Selective Search](#) is used but other methods are appropriate too.
 - For [Semantic Segmentation](#), [Constrained Parametric Min-Cuts](#), CPMC is used.
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