

# 아리랑 위성영상 AI 객체 검출 경진대회

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1 Exploratory Data Analysis

2 Model

Conclusion & Q

Exploratory Data Analysis

Post Processing

Deep Learning Model

Backbone

Head + Neck

STEP 3

STEP 1

STEP 2

Conclusion & QA

https://dacon.i





# Exploratory Data AnalysishSaugmentation



Since there are many orientation variations in aerial images implement the online rotation augmentation. (e.g. 180, 270)







Fig.1 origin rotated Fig.2 rotated Fig.3 rotated



# Exploratory Data Analysisata Augmentation

This allows for the model to learn how to identify objects at a smaller scale than normal. It also encourages the model to localize different types of images in different portions of the frame.



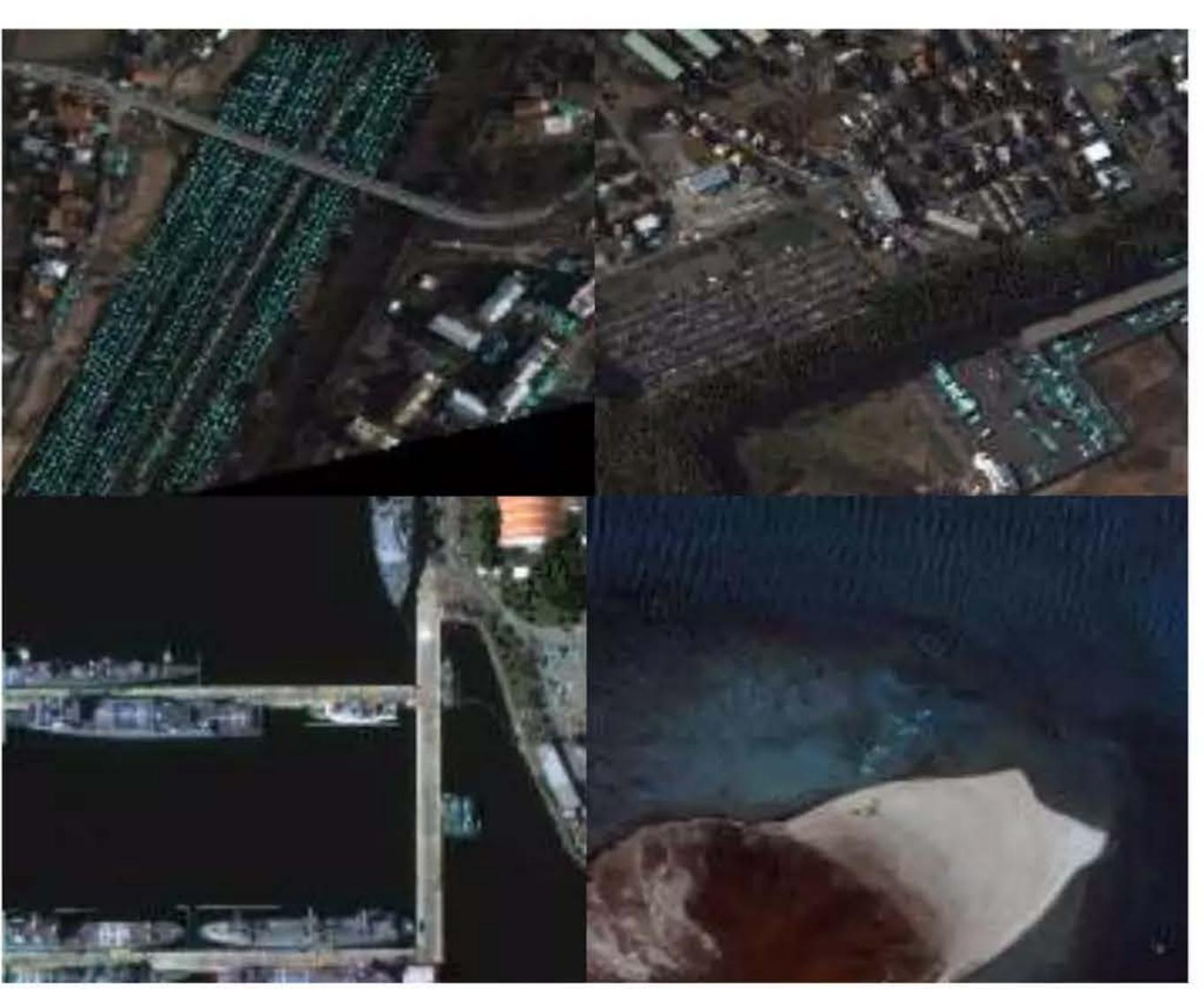


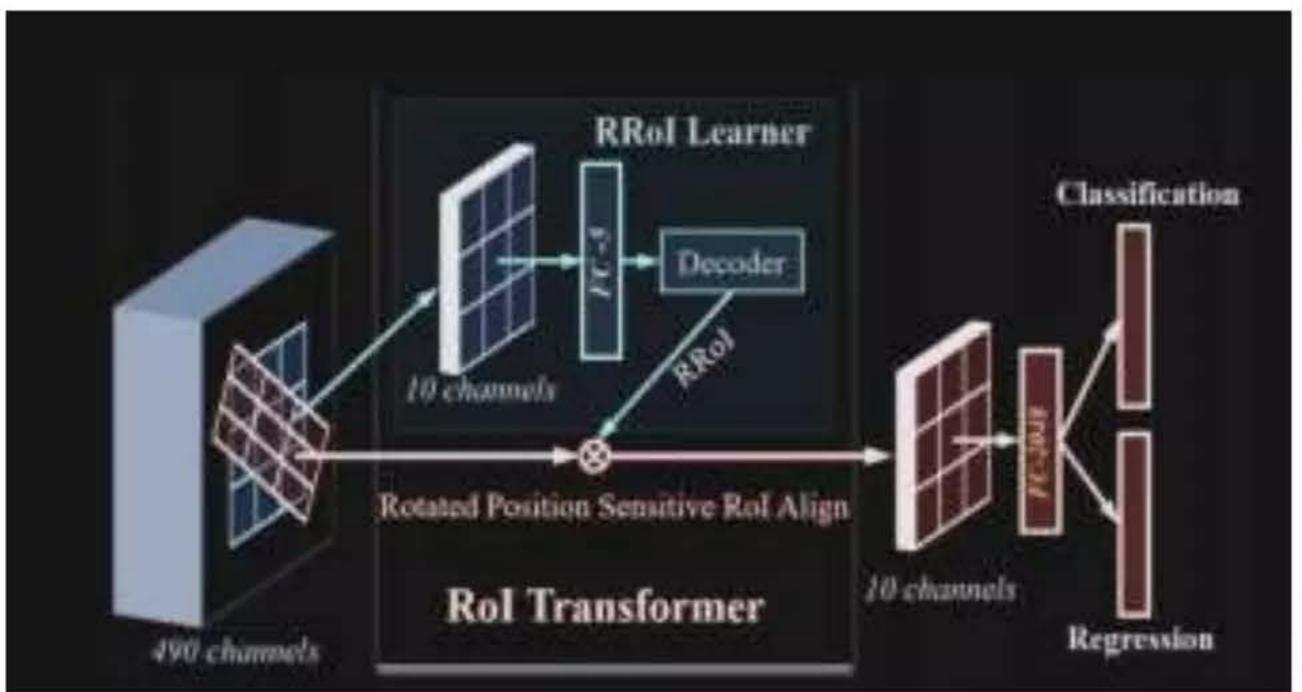
Fig.1 origin image

Fig.2 combines 4 images



#### 2. Model





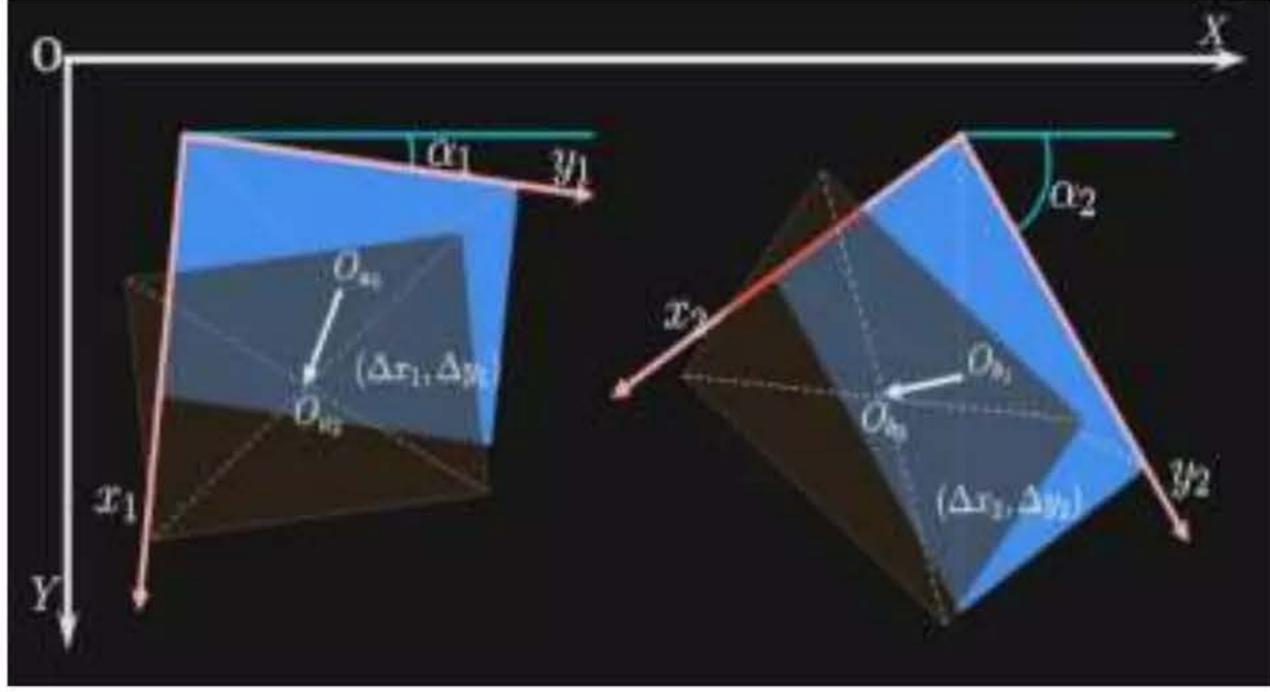
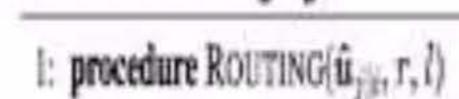


Fig.2 ROI Transform

Fig.1 ROI Transform

Procedure 1 Routing algorithm.



- for all capsule i in layer l and capsule j in layer (l+1):  $b_{ij} \leftarrow 0$ .
- for r iterations do
- for all capsule i in layer  $l: c_i \leftarrow softmax(b_i)$ > softmax computes Eq. 3
- for all capsule j in layer (i+1):  $s_j \leftarrow \sum_i c_{ij} \hat{\mathbf{u}}_{ji}$
- for all capsule j in layer (l+1):  $v_j \leftarrow squash(s_j)$ o squash computes Eq. 1
- for all capsule i in layer l and capsule j in layer (l+1):  $b_{ij} \leftarrow b_{ij} + \hat{\mathbf{u}}_{j|i}.\mathbf{v}_j$ return v

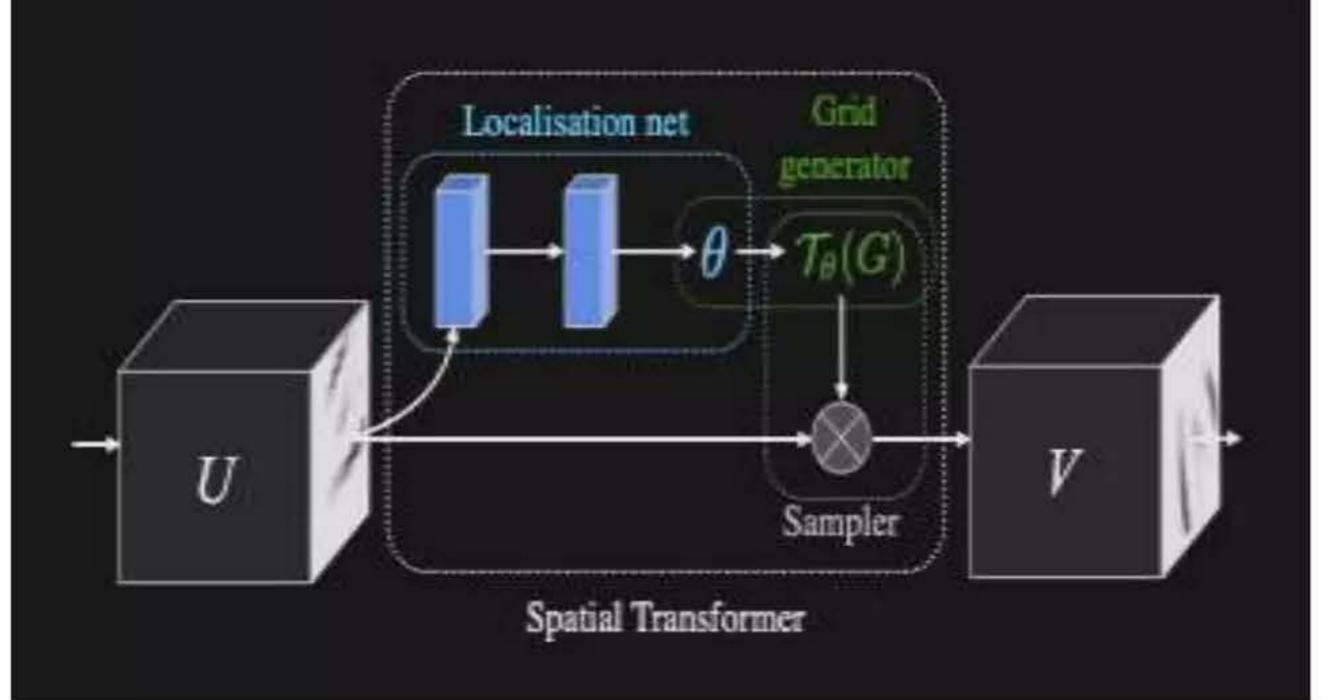
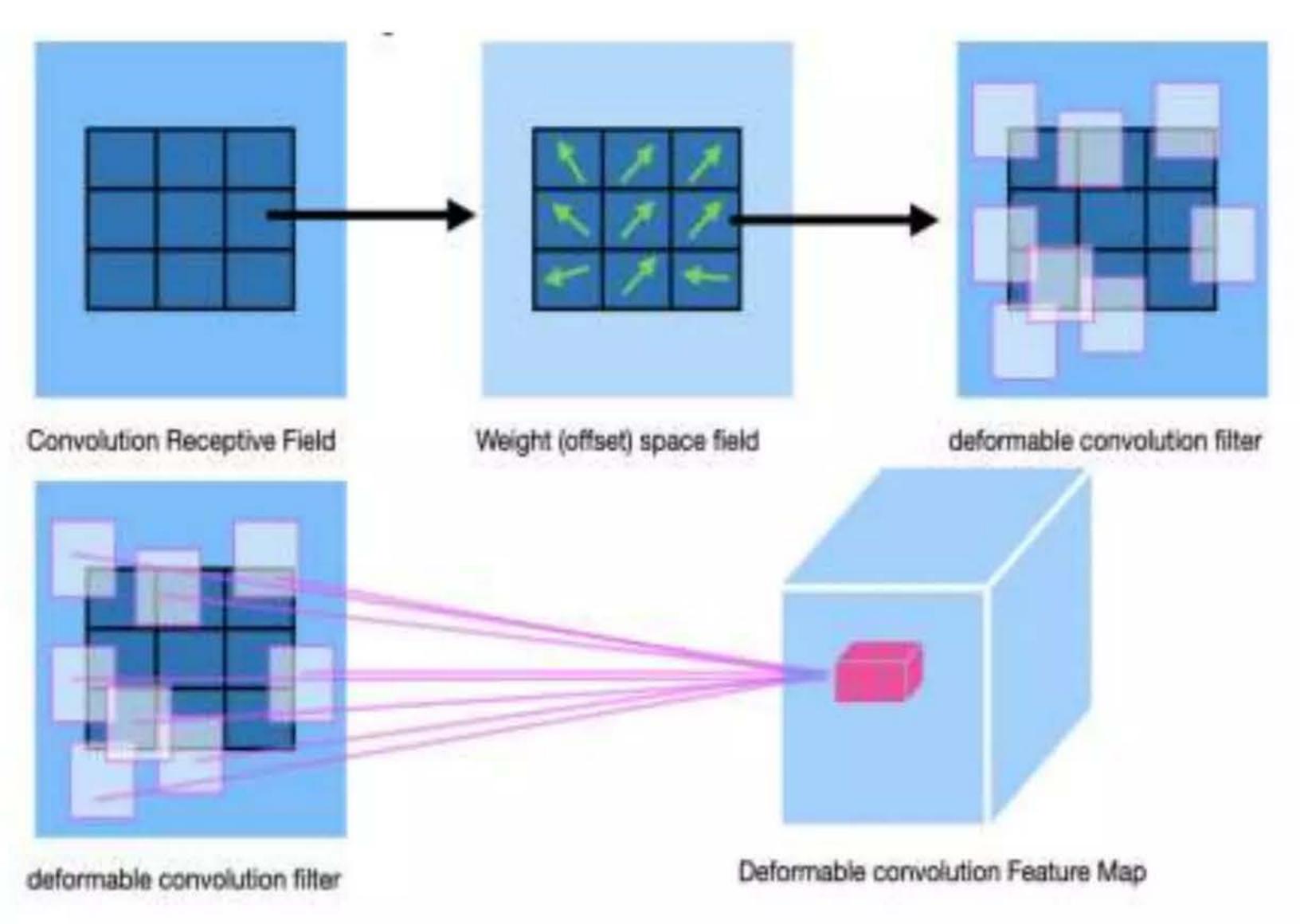


Fig.3 Spatial Transformer Network https://dacon.io

Fig.4 Dynamic Routing Algorithm

### 2. Deformable Convolution Networks





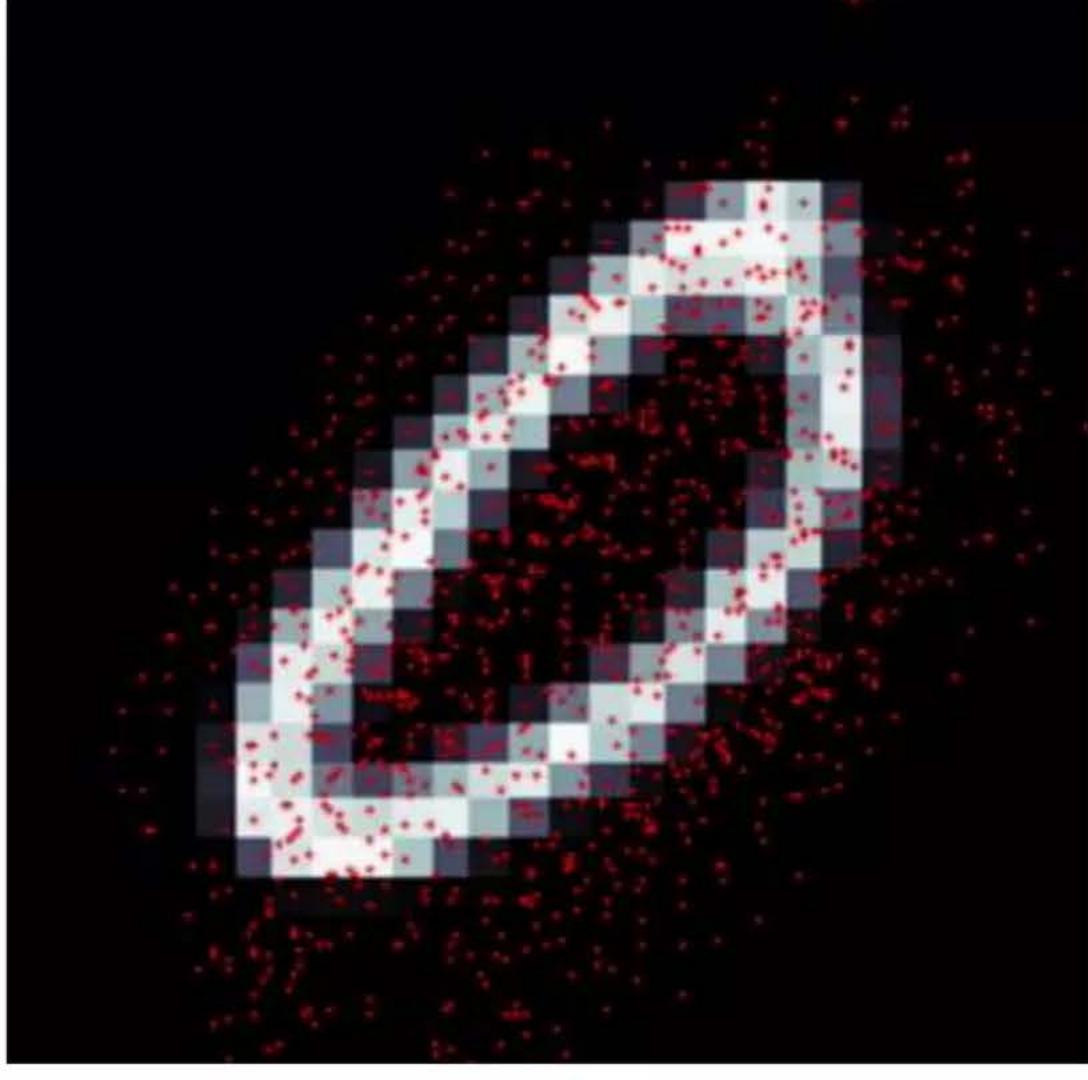


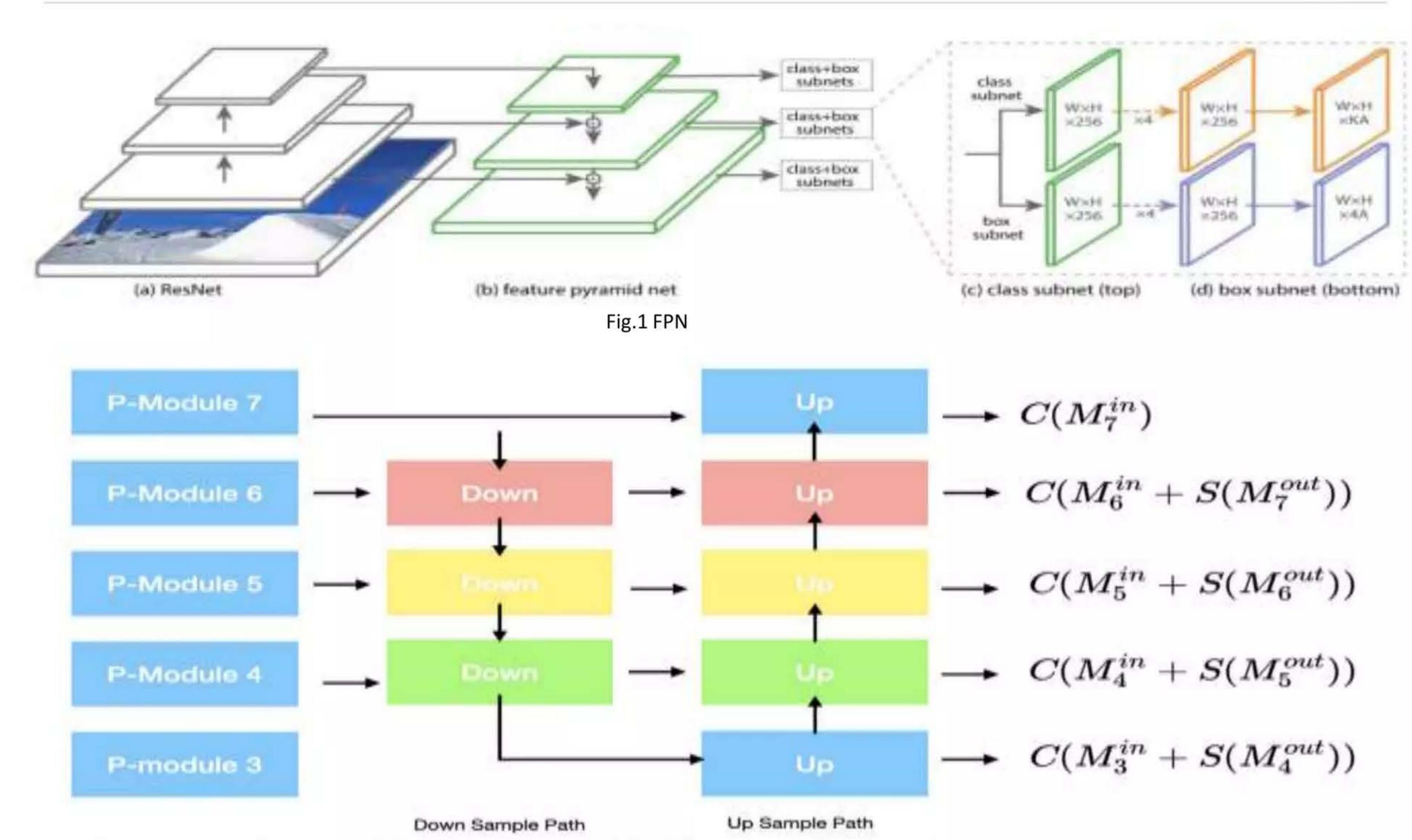
Fig.1 Training DCN Module

Fig.2 DCN sampling locations (from the learned offset)

- 1. Enabling effective modeling of spatial transformation in convolution neural network
- 2. No additional supervision for learning spatial transformation
- 3. Significant accuracy improvements on sophisticated vision tasks

#### 2. Model

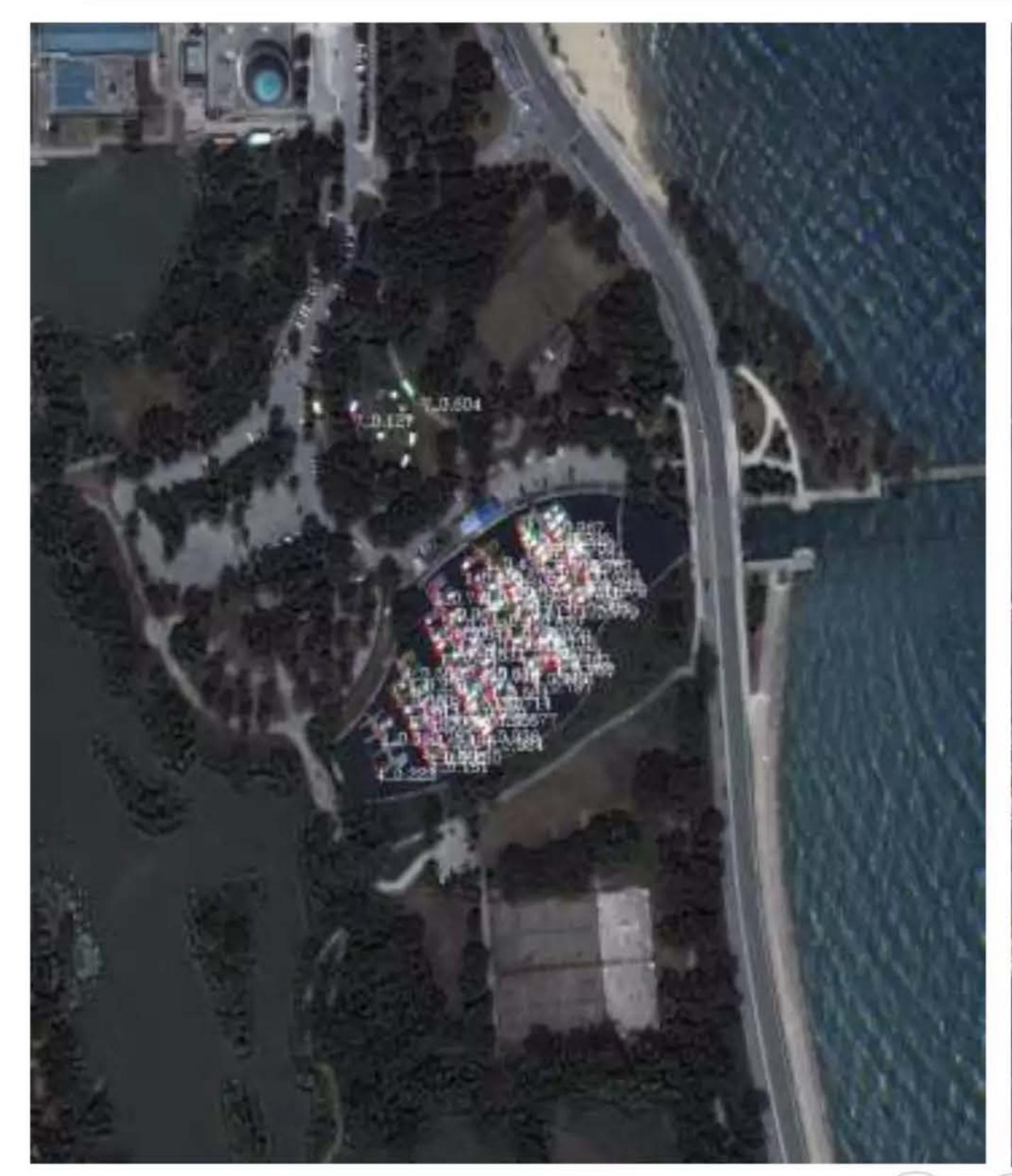


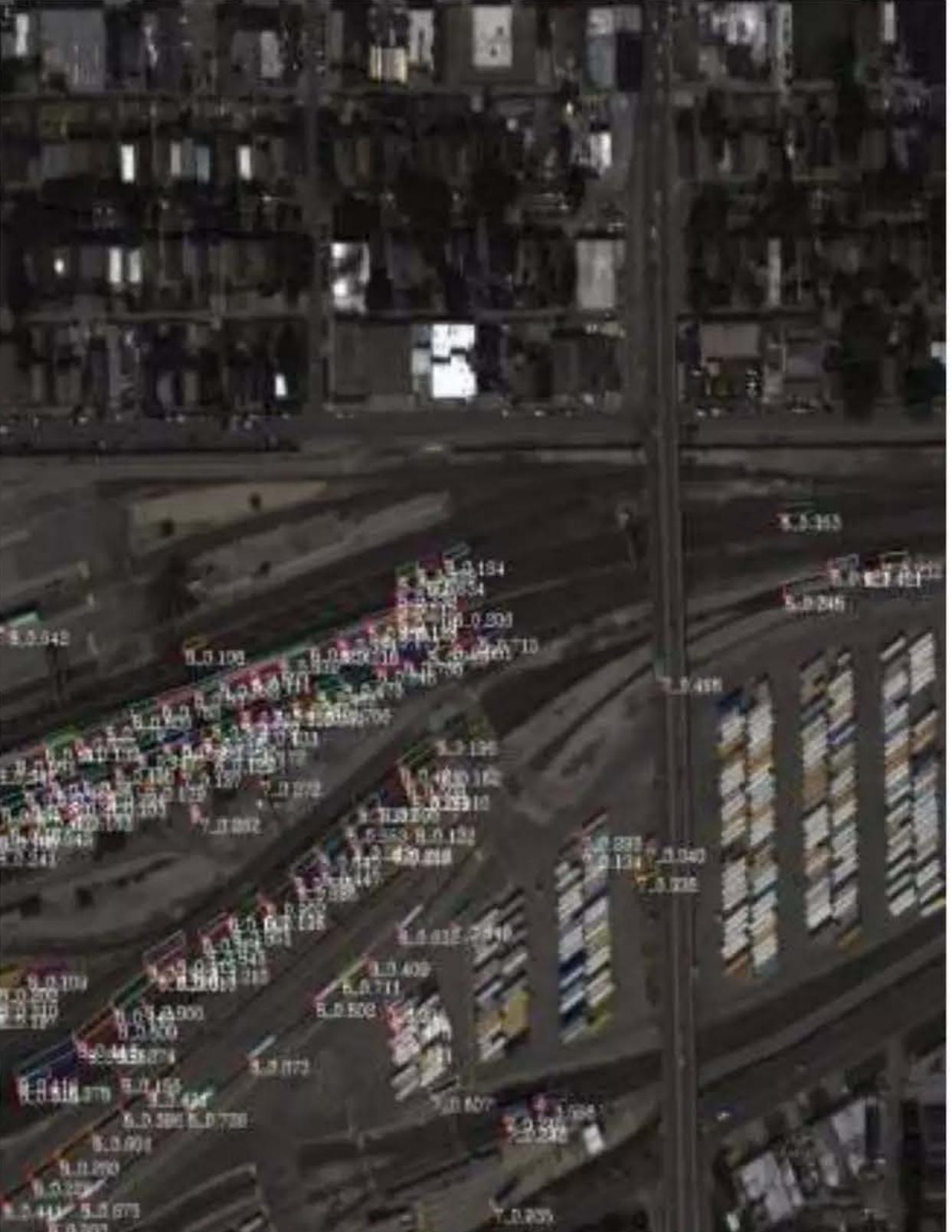


#### Feature Pyramid Module with Pyramid Block

## 3. Result

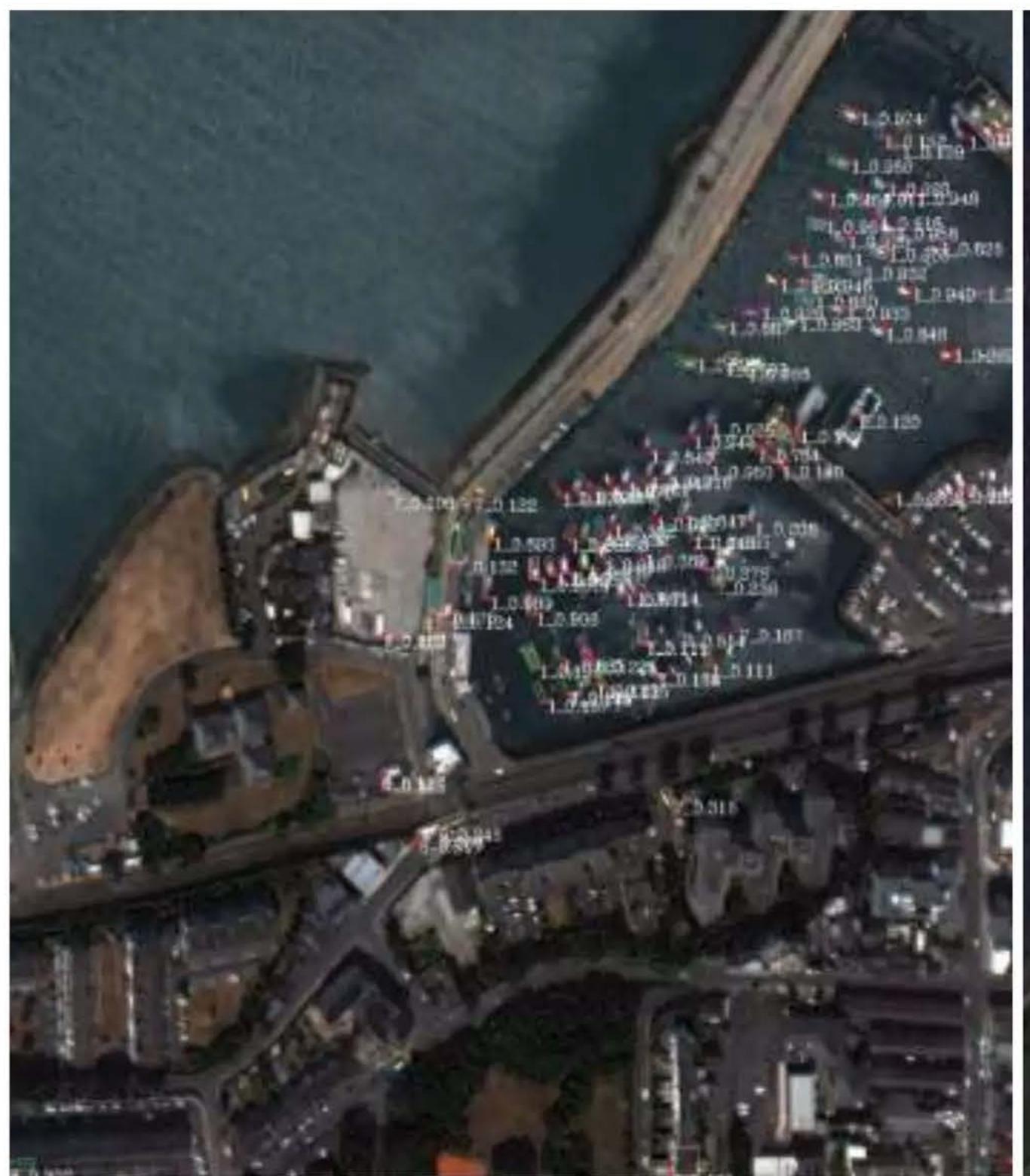






## 3. Result

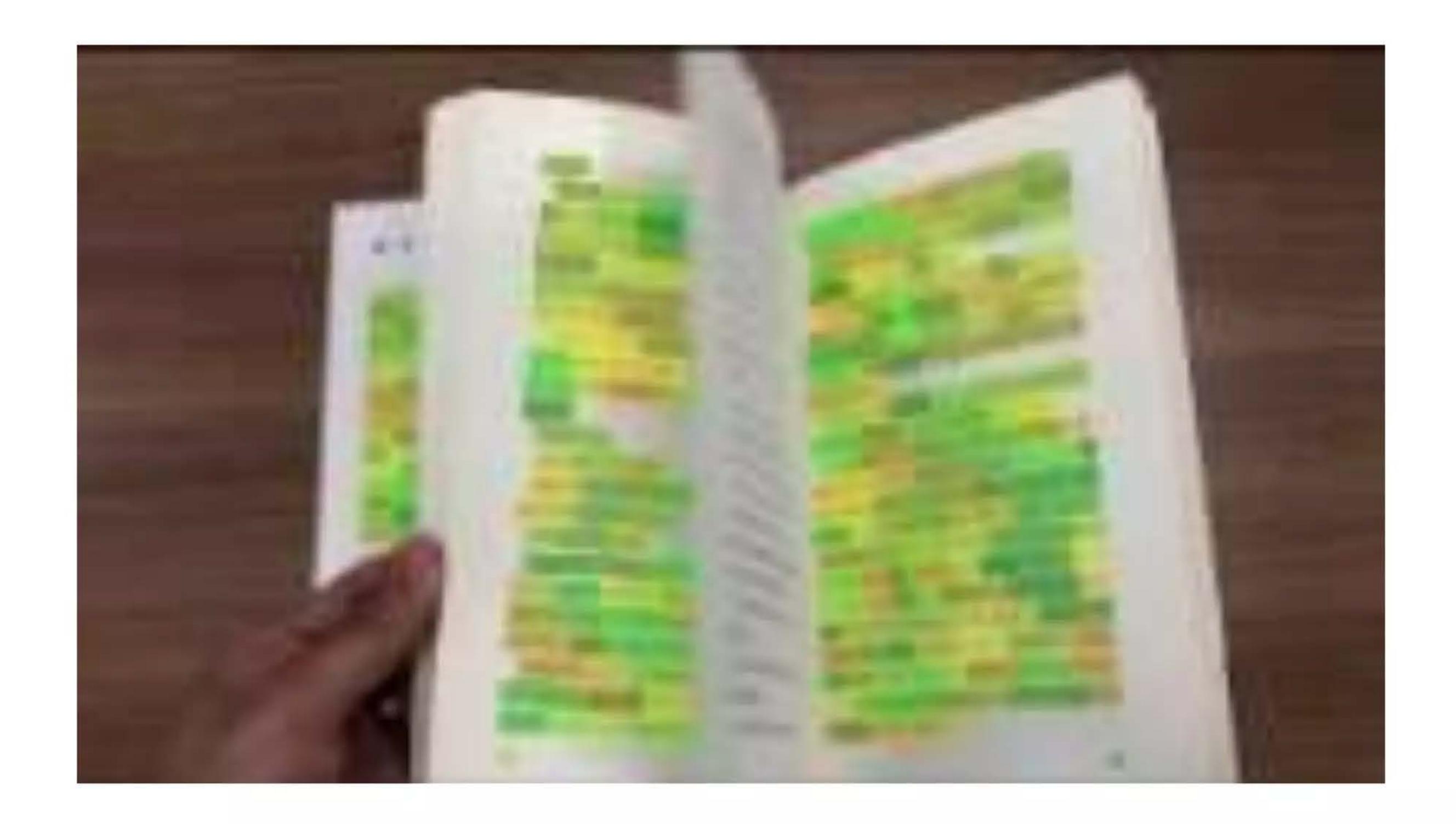






## 3. Attention Expansion Pyramid Network





### 3. Conclusion



- 1. Mean Average Precision
- 2. Align Deep Feature & Activation MAP
- 3. Explainable Neural Networks
- 4. Convolutional vs Transformers



Need good system 1 functionality to make a system efficient

### 4. Reference



- 1. Faster R-CNN: Towards Real-Time Object Detection with Region Proposal Networks
- 2. Fully Convolutional Networks for Semantic Segmentation
- 3. Focal Loss for Dense Object Detection
- 4. Deformable Convolutional Networks
- 5. Learning Rol Transformer for Detecting Oriented Objects in Aerial Images
- 6. YOLOv4: Optimal Speed and Accuracy of Object Detection
- 7. Arbitrary-Oriented Scene Text Detection via Rotation Proposals



