

创建时间: 2019/9/7 11:11

更新时间: 2019/9/7 23:28

作者: Min Xia

URL: <https://zhuanlan.zhihu.com/p/70771042>

---

## NMS

[details information](#)

- [soft NMS](#),
- [Softer NMS](#)

```
Input :  $B = \{b_1, \dots, b_N\}$ ,  $S = \{s_1, \dots, s_N\}$ ,  $N_t$   
         $B$  is the list of initial detection boxes  
         $S$  contains corresponding detection scores  
         $N_t$  is the NMS threshold  
  
begin  
     $\mathcal{D} \leftarrow \{\}$   
    while  $B \neq \text{empty}$  do  
         $m \leftarrow \operatorname{argmax} S$   
         $\mathcal{M} \leftarrow b_m$   
         $\mathcal{D} \leftarrow \mathcal{D} \cup \mathcal{M}$ ;  $B \leftarrow B - \mathcal{M}$   
        for  $b_i$  in  $B$  do  
            if  $\operatorname{iou}(\mathcal{M}, b_i) \geq N_t$  then  
                 $B \leftarrow B - b_i$ ;  $S \leftarrow S - s_i$   
            end NMS  
             $s_i \leftarrow s_i f(\operatorname{iou}(\mathcal{M}, b_i))$  Soft-NMS  
        end  
    end  
    return  $\mathcal{D}, S$   
end
```

---

**Algorithm 1** softer-NMS

---

$\mathcal{B}$  is  $N \times 4$  matrix of initial detection boxes.  $\mathcal{S}$  contains corresponding detection scores.  $\mathcal{C}$  is  $N \times 4$  matrix of corresponding variances.  $N_t$  is the softer NMS threshold. The lines in blue and in green are soft-NMS and softer-NMS respectively.

```
 $\mathcal{B} = \{b_1, \dots, b_N\}, \mathcal{S} = \{s_1, \dots, s_N\}, \mathcal{C} = \{\sigma_1^2, \dots, \sigma_N^2\}, N_t$   
 $\mathcal{D} \leftarrow \{\}$   
 $\mathcal{T} \leftarrow \mathcal{B}$   
while  $\mathcal{T} \neq \text{empty}$  do  
   $m \leftarrow \text{argmax } \mathcal{S}$   
   $\mathcal{M} \leftarrow b_m$   
   $\mathcal{T} \leftarrow \mathcal{T} - \mathcal{M}$   
   $\mathcal{S} \leftarrow \mathcal{S} f(\text{IoU}(\mathcal{M}, \mathcal{T}))$  ▷ soft-NMS  
   $idx \leftarrow \text{IoU}(\mathcal{M}, \mathcal{B}) \geq N_t$  ▷ softer-NMS  
   $\mathcal{M} \leftarrow \mathcal{B}[idx] / \mathcal{C}[idx] / \text{sum}(1 / \mathcal{C}[idx])$   
   $\mathcal{D} \leftarrow \mathcal{D} \cup \mathcal{M}$   
end while  
return  $\mathcal{D}, \mathcal{S}$ 
```

---

[soft nms implement](#)

- [IoU Guided NMS](#)

---

**Algorithm 1** IoU-guided NMS. Classification confidence and localization confidence are disentangled in the algorithm. We use the localization confidence (the predicted IoU) to rank all detected bounding boxes, and update the classification confidence based on a clustering-like rule.

---

**Input:**  $\mathcal{B} = \{b_1, \dots, b_n\}$ ,  $\mathcal{S}$ ,  $\mathcal{I}$ ,  $\Omega_{\text{nms}}$

$\mathcal{B}$  is a set of detected bounding boxes.

$\mathcal{S}$  and  $\mathcal{I}$  are functions (neural networks) mapping bounding boxes to their classification confidence and IoU estimation (localization confidence) respectively.

$\Omega_{\text{nms}}$  is the NMS threshold.

**Output:**  $\mathcal{D}$ , the set of detected bounding boxes with classification scores.

```
1:  $\mathcal{D} \leftarrow \emptyset$   
2: while  $\mathcal{B} \neq \emptyset$  do  
3:    $b_m \leftarrow \text{arg max } \mathcal{I}(b_j)$   
4:    $\mathcal{B} \leftarrow \mathcal{B} \setminus \{b_m\}$   
5:    $s \leftarrow \mathcal{S}(b_m)$   
6:   for  $b_j \in \mathcal{B}$  do  
7:     if  $\text{IoU}(b_m, b_j) > \Omega_{\text{nms}}$  then  
8:        $s \leftarrow \max(s, \mathcal{S}(b_j))$   
9:        $\mathcal{B} \leftarrow \mathcal{B} \setminus \{b_j\}$   
10:    end if  
11:  end for  
12:   $\mathcal{D} \leftarrow \mathcal{D} \cup \{(b_m, s)\}$   
13: end while  
14: return  $\mathcal{D}$ 
```

---

知乎 @橙煦猿

- [Conv NMS](#)
- [Learning NMS](#)

- Adaptive NMS

**Input** :  $\mathcal{B} = \{b_1, \dots, b_N\}$ ,  $\mathcal{S} = \{s_1, \dots, s_N\}$ ,  
 $\mathcal{D} = \{d_1, \dots, d_N\}$ ,  $N_t$   
 $\mathcal{B}$  is the list of initial detection boxes  
 $\mathcal{S}$  contains corresponding detection scores  
 $\mathcal{D}$  contains corresponding detection densities  
 $N_t$  is the NMS threshold

```

begin
   $\mathcal{F} \leftarrow \{\}$ 
  while  $\mathcal{B} \neq \text{empty}$  do
     $m \leftarrow \text{argmax } \mathcal{S}$ 
     $\mathcal{M} \leftarrow b_m$ 
     $N_{\mathcal{M}} \leftarrow \max(N_t, d_m)$ 
     $\mathcal{F} \leftarrow \mathcal{F} \cup \mathcal{M}$ ;  $\mathcal{B} \leftarrow \mathcal{B} - \mathcal{M}$ 
    for  $b_i$  in  $\mathcal{B}$  do
      if  $\text{iou}(\mathcal{M}, b_i) \geq N_t$  then
         $\mathcal{B} \leftarrow \mathcal{B} - b_i$ ;  $\mathcal{S} \leftarrow \mathcal{S} - s_i$ ;
      end
    end
  end
  return  $\mathcal{F}, \mathcal{S}$ 
end

```

if  $\text{iou}(\mathcal{M}, b_i) \geq N_t$  then  
 |  $\mathcal{B} \leftarrow \mathcal{B} - b_i$ ;  $\mathcal{S} \leftarrow \mathcal{S} - s_i$ ;  
 end  
 Greedy-NMS

if  $\text{iou}(\mathcal{M}, b_i) \geq N_{\mathcal{M}}$  then  
 |  $\mathcal{B} \leftarrow \mathcal{B} - b_i$ ;  $\mathcal{S} \leftarrow \mathcal{S} - s_i$ ;  
 end  
 Adaptive-NMS

- [Pure NMS Network](#)
- Yes-Net: An effective Detector Based on Global Information