

# A Joint Framework Towards Class-aware and Class-agnostic Alignment for Few-shot Segmentation

Kai Huang<sup>1\*</sup>[0000-0003-3700-9052], Mingfei Cheng<sup>2\*</sup>[0000-0002-8982-1483], and Yang Wang<sup>1</sup> and Bochen Wang<sup>1</sup> and Ye Xi<sup>1</sup> and feigege wang<sup>1</sup> and Peng Chen<sup>1</sup>

<sup>1</sup> Alibaba Group, China

{zhouwan.hk, wanyuan.wy, bochen.wbc, yx150449, feigege.wfgg, yuanshang.cp}@alibaba-inc.com

<sup>2</sup> Singapore Management University, Singapore  
snowbirds.mf@gmail.com

In this material, we firstly report the further ablation study on Hybrid Prototype Alignment Module (HPAM) to explore the application of multi-scale, and then further show the effect of block sizes ( $m$ ) and top $k$  ratio ( $k$ ) in the P2P module.

## 1 More Study on HPAM with Multi-Scale Scheme

Table 1 reports the ablation studies of multi-scale scheme. “BL” means the baseline of this ablation study which removes the multi-scale scheme as well as the Hybrid Prototype Alignment Module, “BL\*” indicates the baseline with multi-scale scheme. We also conduct the single HPAM with only choosing the middle layer features, which are represented as “BL  $w$ / single HPAM” and “BL\*  $w$ / single HPAM” respectively. And “BL\*  $w$ / HPAM” is the complete framework of our JC<sup>2</sup>A, in which the HPAM acts in each scale feature maps. From Table 1, we can draw the following conclusion as: 1) It is clear that multi-scale scheme can greatly improve the segmentation performance which has been verified in many FSS methods. 2) Not only for normal baseline but also the baseline with multi-scale scheme, our single HPAM can still make positive effect. 3) the model with multi-scale HPAM achieves highest accuracy in all folds, proving the superiority of our HPAM with multi-scale scheme.

## 2 Block size and top $k$ settings

We further analyzed the effect of different block sizes ( $m$ ) and the choice of top $k$  ratio ( $k$ ) in the P2B. As shown in Table 2, we find that the performance drops with the expansion of block size. These result with different  $m$  settings indicate that the large block is able to introduce more irrelevant semantic noises which weaken the representation of target parts within specific regions. Besides, setting  $k$  to 20% achieves the best performance while using all blocked features brings

---

\* Equal contribution.

**Table 1.** The performance comparison of 1-shot and 5-shot segmentation on PASCAL-5<sup>t</sup>. Best-performing results are highlighted in bold.

Method	1-shot						5-shot					
	fold-0	fold-1	fold-2	fold-3	<i>mIoU</i>	FB-IoU	fold-0	fold-1	fold-2	fold-3	<i>mIoU</i>	FB-IoU
BL	59.3	67.0	52.9	53.5	58.2	69.5	62.8	67.3	51.6	55.7	59.4	70.2
BL*	62.8	69.6	54.8	56.5	60.9	72.2	65.8	68.7	54.9	58.9	62.1	72.9
BL <i>w/</i> single HPAM	63.7	69.5	54.0	57.9	61.3	72.7	65.1	69.9	55.4	59.0	62.4	73.0
BL* <i>w/</i> single HPAM	65.4	71.2	56.5	59.0	63.0	75.2	67.7	71.8	57.3	61.4	64.6	74.8
BL* <i>w/</i> HPAM (our)	<b>67.3</b>	<b>72.4</b>	<b>57.7</b>	<b>60.7</b>	<b>64.5</b>	<b>76.5</b>	<b>68.6</b>	<b>72.9</b>	<b>58.7</b>	<b>62.0</b>	<b>65.4</b>	<b>76.8</b>

negative effects. We can infer that our block selection of P2B can keep most relevant class-aware information without reducing computational efficiency.

**Table 2.** Ablation study on different block sizes ( $m$ ) and top $k$  ratio ( $k$ ) settings in P2B. “./.” represents the results of “1-shot/5-shot”.

$k$	10%	20%	30%	40%	50%	60%	70%	80%
$m=3$	62.1/63.7	<b>64.5/65.4</b>	64.3/65.0	64.2/65.1	64.4/65.3	64.5/65.2	64.1/64.9	63.0/64.3
$m=5$	61.4/62.8	63.2/64.1	63.0/64.0	63.5/64.7	62.9/63.4	63.3/63.8	62.2/63.2	62.0/63.7
$m=7$	60.7/61.9	63.0/64.4	62.9/63.7	62.5/63.7	63.1/64.2	62.7/64.4	62.0/63.6	61.6/62.9