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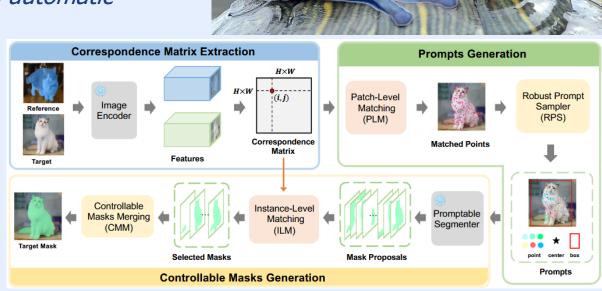






Motivation

- Segment Anything Model (SAM) relies on point/box/mask prompts
 - Cannot apply for automatic semantic segmentation
- Previous methods (Matcher, PerSAM) introduce Few-shot
 Semantic Segmentation (FSS) for SAM-based automatic
 semantic segmentation.
 - Lots of external hyperparameters
 - Low efficiency
 - Iterative mask generation



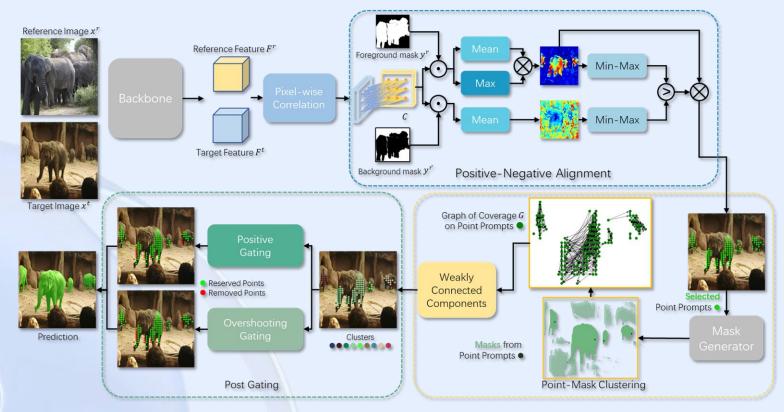
^[1] Kirillov, A., Mintun, E., Ravi, N., Mao, H., Rolland, C., Gustafson, L., ... & Girshick, R. (2023). Segment anything. In *Proceedings of the IEEE/CVF International Conference on Computer Vision* (pp. 4015-4026). [2] Zhang, R., Jiang, Z., Guo, Z., Yan, S., Pan, J., Ma, X., ... & Li, H. (2023). Personalize segment anything model with one shot. arXiv preprint arXiv:2305.03048.

^[3] Liu, Y., Zhu, M., Li, H., Chen, H., Wang, X., & Shen, C. (2023). Matcher: Segment anything with one shot using all-purpose feature matching. arXiv preprint arXiv:2305.13310.

Our Contribution

- ✓ Training-Free
- ✓ External-Hyperparameter-Free
- ✓ No iterative mask generation

- ✓ Fast inference within 2s per image*
- ✓ New state-of-the-art performance
- ✓ Effective in various domains

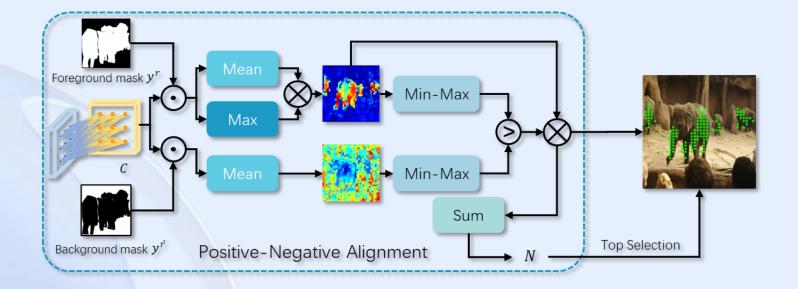


^{*} Evaluated on a single NVIDIA RTX2080Ti

Positive-Negative Alignment

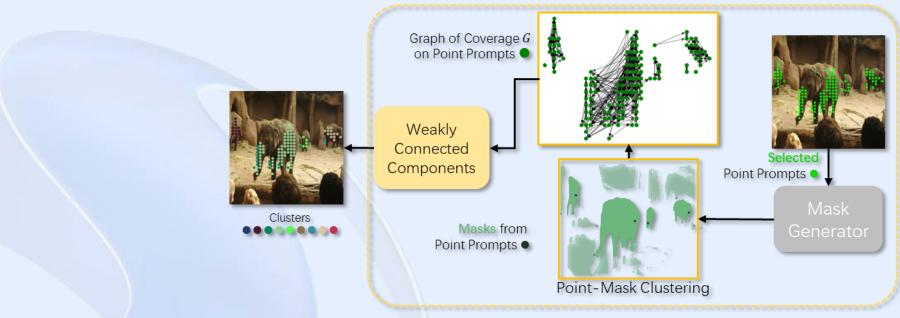
- \triangleright C: Pixel-wise correlation ($hw \times hw$) between reference and target features.
- Mean/Max: Get mean/max value on the dimension corresponding to reference

- features and generate similarity maps.
- Min-Max: Highlight the similarity values of foreground/background.



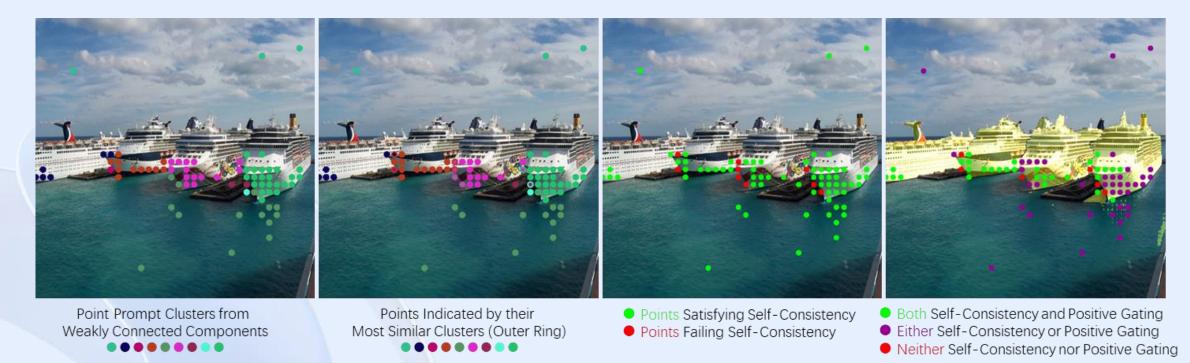
Point-Mask Clustering

- Each point P_l corresponds to a unique mask \hat{y}_l from SAM.
- Selected Points from fine-grained features need to align to masks from coarsegrained features.
- PMC module constructs directed graph G
 according to the coverage of masks over
 other points.
 - Each weakly connected component become a cluster.

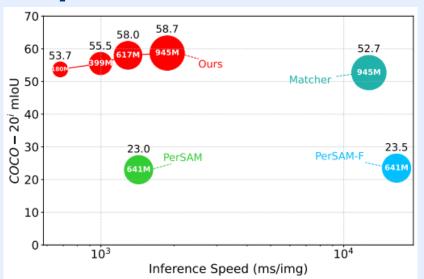


Post Gating

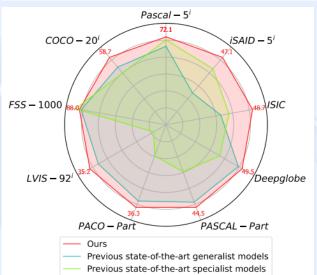
- Positive Gating: Compare number of positive and negative pixels according to positive and negative similarity maps.
- Overshooting Gating: Retain points having highest similarity on regions of its corresponding mask (with distance factor).



Experiments



| Methods | Pascal-5i | | COCO-20i | | FSS-1000 | | LVIS-92i | |
|--------------------------|-----------|--------|----------|--------|----------|--------|----------|--------|
| Wiethous | 1-shot | 5-shot | 1-shot | 5-shot | 1-shot | 5-shot | 1-shot | 5-shot |
| specialist model | | | | | | | | |
| HSNet [14][CVPR21] | 66.2 | 70.4 | 41.2 | 49.5 | 86.5 | 88.5 | 17.4 | 22.9 |
| VAT [50][ECCV22] | 67.9 | 72.0 | 41.3 | 47.9 | 90.3 | 90.8 | 18.5 | 22.7 |
| HDMNet [44][CVPR23] | 69.4 | 71.8 | 50.0 | 56.0 | - | - | - | - |
| AMFormer [45][NeurIPS23] | 70.7 | 73.6 | 51.0 | 57.3 | - | - | - | - |
| generalist model | | | | | | | | |
| PerSAM [12][ICLR24] | 43.1 | - | 23.0 | - | 71.2 | - | 11.5 | - |
| PerSAM-F [12][ICLR24] | 48.5 | - | 23.5 | - | 75.6 | - | 12.3 | - |
| Matcher [13][ICLR24] | 68.1 | 74.0 | 52.7 | 60.7 | 87.0 | 89.6 | 33.0 | 40.0 |
| VRP-SAM [49][CVPR24] | 71.9 | - | 53.9 | - | - | - | - | - |
| Ours | 72.1 | 82.6 | 58.7 | 66.8 | 88.0 | 88.9 | 35.2 | 44.2 |



| | One-shot Part Seg. | | Cross Domain FSS | | | | | |
|-----------------------|--------------------|-----------|------------------|--------|--------|--------|--------|------------------|
| Methods | PASCAL-Part | PACO-Part | Deep | globe | IS | IC | iSAI | D-5 ⁱ |
| | 1-shot | 1-shot | 1-shot | 5-shot | 1-shot | 5-shot | 1-shot | 5-shot |
| specialist model | | | | | | | | |
| HSNet [14][CVPR21] | 32.4 | 22.6 | 29.7 | 35.1 | 31.2 | 35.1 | 34.1 | 40.4 |
| DRA [51][CVPR24] | - | - | 41.3 | 50.1 | 40.8 | 48.9 | - | - |
| FRINet [52][TGRS23] | - | - | - | - | - | - | 42.6 | 44.5 |
| generalist model | | | | | | | | |
| PerSAM [12][ICLR24] | 32.5 | 22.5 | 31.4 | - | 23.9 | - | 19.2 | - |
| PerSAM-F [12][ICLR24] | 32.9 | 22.7 | 35.0 | - | 23.6 | - | 20.3 | - |
| Matcher [13][ICLR24] | 42.9 | 34.7 | 48.1 | 50.9 | 38.6 | 35.0 | 33.3 | 34.3 |
| Ours | 44.5 | 36.3 | 49.5 | 57.7 | 48.7 | 55.2 | 47.1 | 52.4 |

Ablation Studies

Table 3: Ablation study of Point Selection.

| S_{mean}^+ | S_{max}^+ | S_{mean}^- | Top N | mIoU |
|--------------|--------------|--------------|---------|------|
| ✓ | | ✓ | ✓ | 53.1 |
| | ✓ | ✓ | ✓ | 54.1 |
| ✓ | \checkmark | ✓ | | 56.4 |
| \checkmark | \checkmark | | ✓ | 51.5 |
| \checkmark | \checkmark | ✓ | ✓ | 58.7 |

Table 4: Ablation study of PMC and Post-Gating.

| PG | | OG | | COCO-20i | LVIS-92i |
|--------|-------|--------|------|----------|----------|
| Strong | Weak | Strong | Weak | | |
| | | | | 44.0 | 24.2 |
| | ✓ | | | 57.1 | 34.3 |
| ✓ | | | | 57.1 | 33.9 |
| | ✓ | ✓ | | 56.7 | 35.2 |
| | ✓ | | ✓ | 58.7 | 35.2 |
| | k-mea | ans++ | | 57.5 | 34.0 |

Table 5: Ablation study of positive gating on each cluster. M.G. represents the Mask Growth algorithm.

| Strategies | M.G. | COCO-20i | PASCAL-Part |
|------------|----------|---------------------|--------------|
| Sum | ✓ | 55.3 58.6 | 39.1 44.3 |
| Num | √ | 57.1 58.7 | 42.2 44.5 |

Table 6: Ablation on the strategies of Self-Consistency measurement.

| Strategies | mIoU | Δ |
|--------------------|------|------|
| None | 57.1 | 0.0 |
| Point Sim. | 56.7 | -0.4 |
| MAP Sim. | 57.7 | +0.6 |
| Mean Sim. W/o dist | 49.1 | -8.0 |
| Mean Sim. (Ours) | 58.7 | +1.6 |

Visualization



