

Relative Position and Map Networks in Few-shot Learning for Image Classification

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UESTC

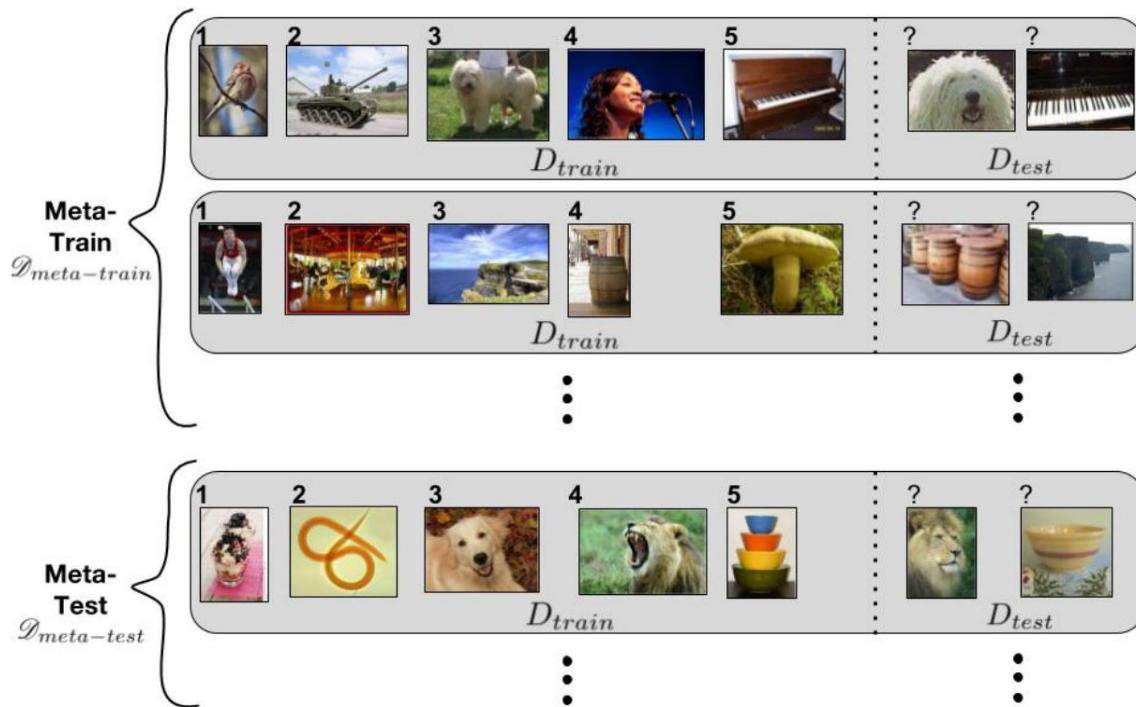


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Our codes will be released at <https://github.com/chrisyxue/RMN-RPN-for-FSL>.

Introduction



Ravi, S., & Larochelle, H. (2016). Optimization as a model for few-shot learning.

Baseline

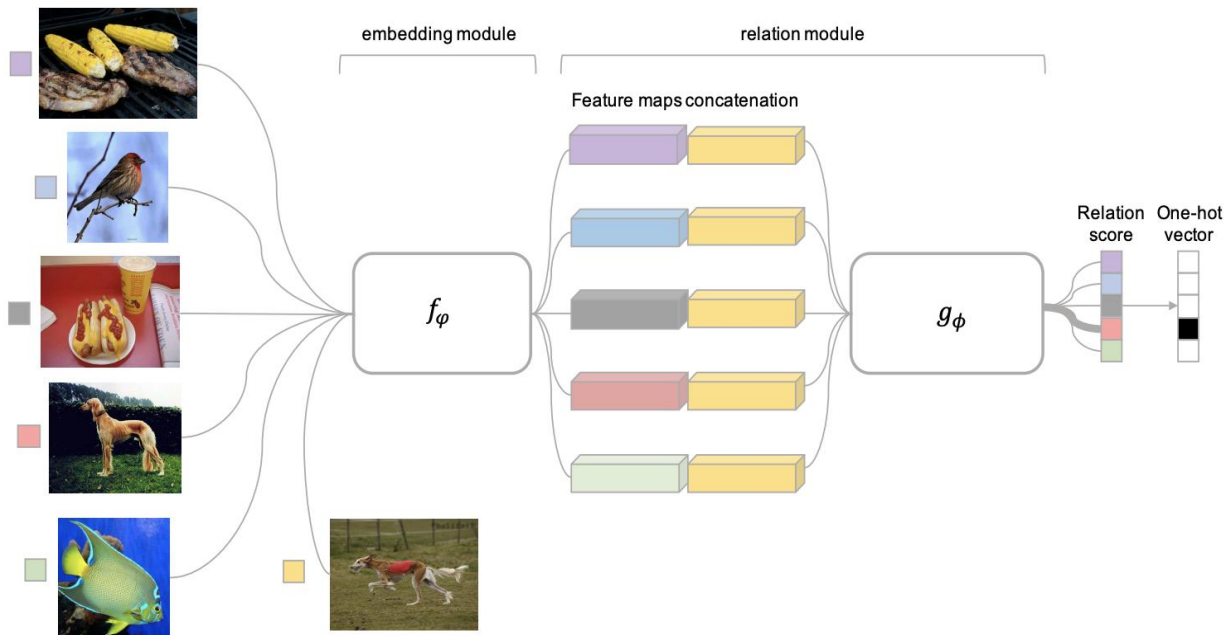
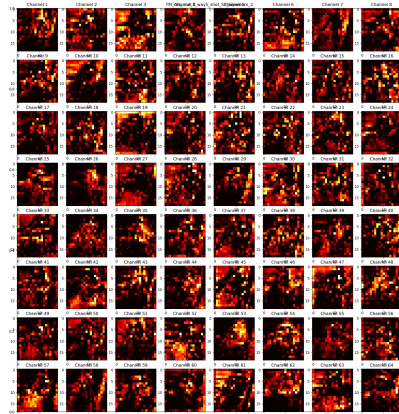


Figure 1: Relation Network architecture for a 5-way 1-shot problem with one query example.

Sung, Flood, et al. "Learning to compare: Relation network for few-shot learning." Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition. 2018.

Motivation

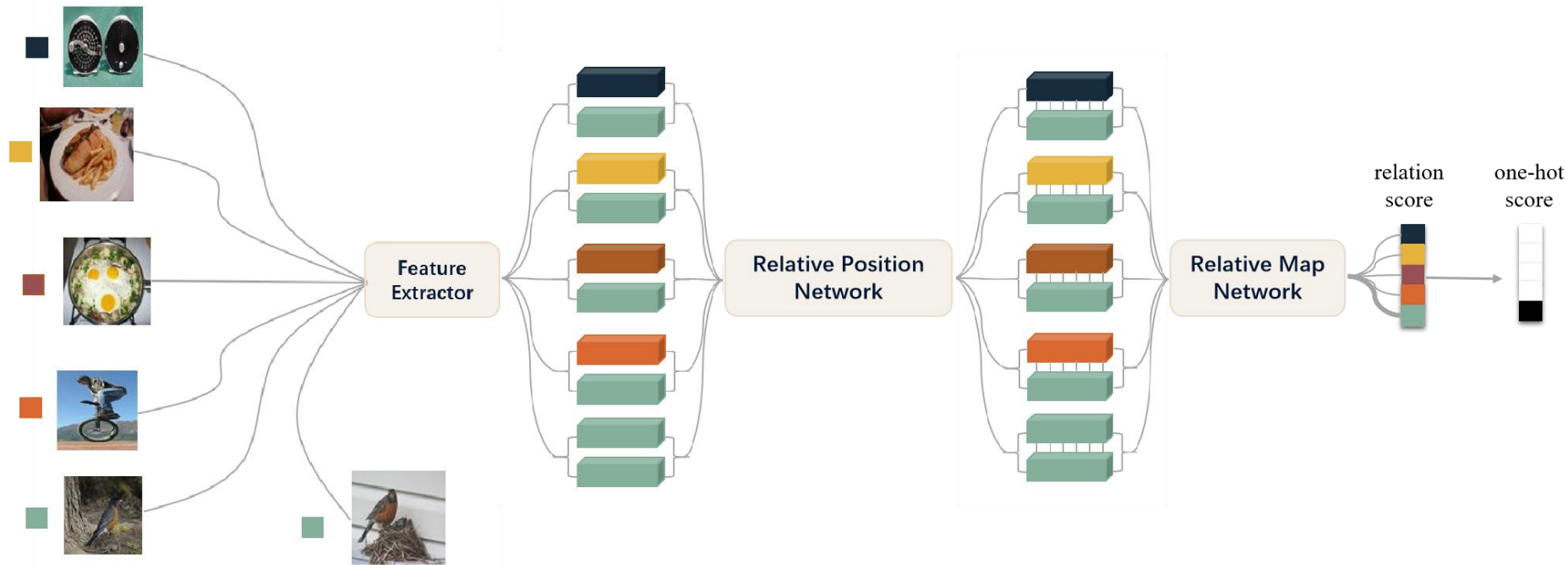


RMN: different channels have different descriptions

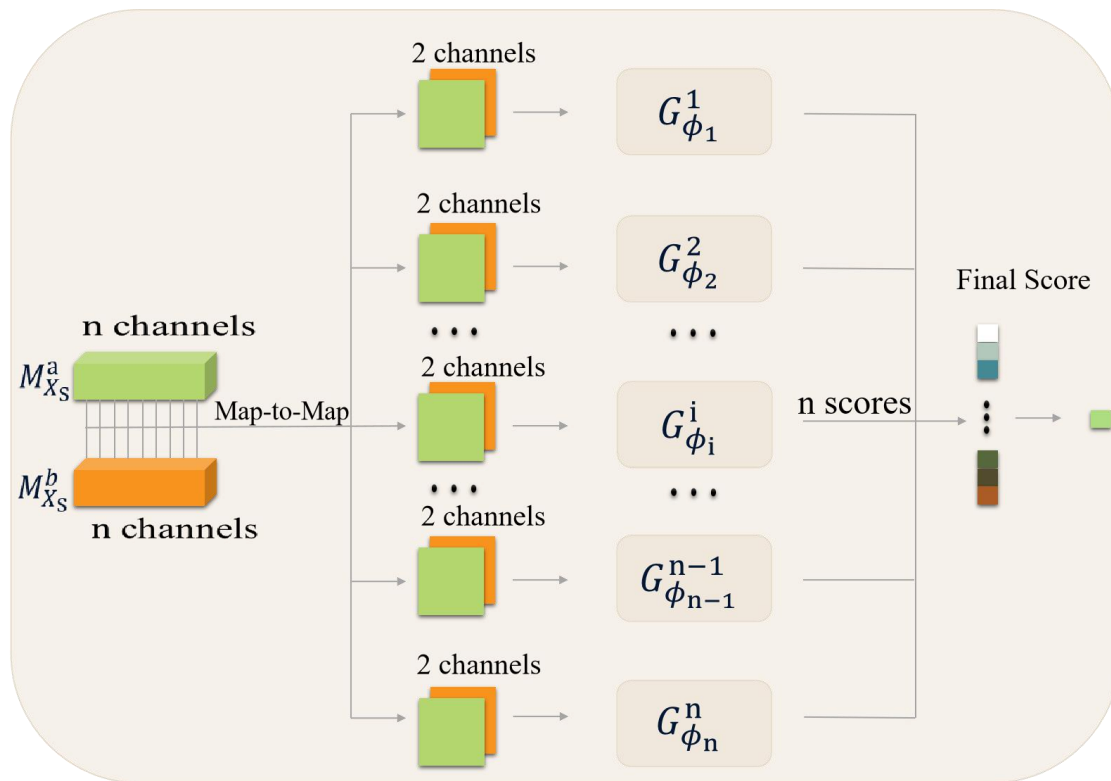


RPN: the importance of each position is different

Architecture

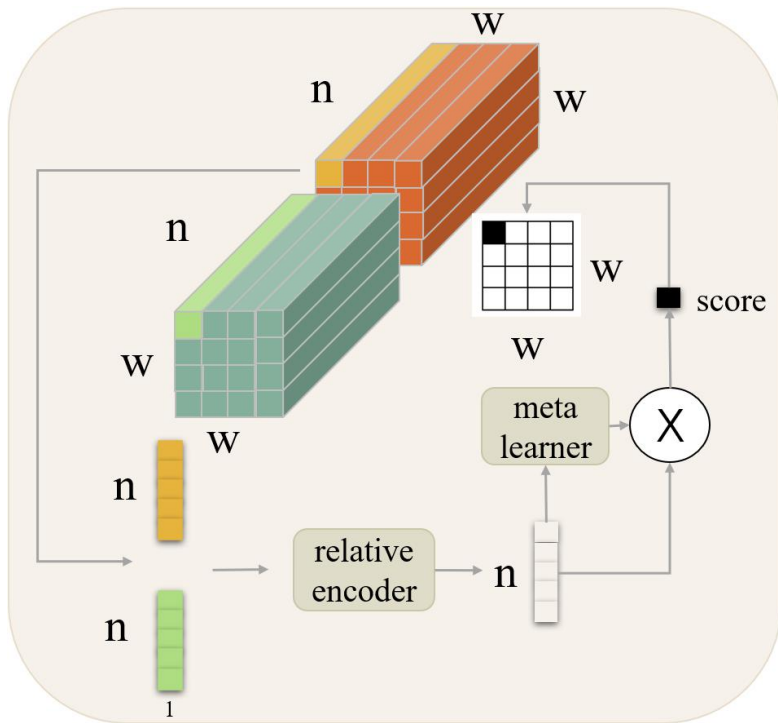


Relative Map Network



$$P_{S,Q} = \text{Sig}\left(\sum_{i=1}^n w_i G_{\phi_i}^i(M_{x_S}^i, M_{x_Q}^i)\right)$$

Relative Position Network



$$V_{i,j}^{s,q} = H([v_{i,j}^S, v_{i,j}^Q])$$

$$w = W_2 \cdot \sigma(W_1 \cdot V_{i,j}^{s,q})$$

$$Att_{i,j} = w^T V_{i,j}^{s,q}$$

$$M_{x^Q} := M_{x^Q} + Att \otimes M_{x^Q}$$

Experiments



Table 1. Mean accuracies (%) of different methods on the MiniImageNet dataset. Results are obtained over 600 test episodes with 95% confidence intervals.

| Model | MiniImageNet (5-way) | |
|---------------------------|-----------------------------------|-----------------------------------|
| | 1-shot | 5-shot |
| MATCHING NETS [21] | 43.56 \pm 0.84 | 55.31 \pm 0.73 |
| META LSTM [15] | 43.44 \pm 0.77 | 60.60 \pm 0.71 |
| MAML [3] | 48.70 \pm 1.84 | 63.11 \pm 0.92 |
| PROTOTYPICAL NETS [19] | 49.42 \pm 0.78 | 68.20 \pm 0.66 |
| META SGD [12] | 50.47 \pm 1.87 | 64.03 \pm 0.94 |
| RN [20] | 50.44 \pm 0.82 | 65.32 \pm 0.70 |
| GNN [17] | 50.33 \pm 0.36 | 66.41 \pm 0.63 |
| PABN [6] | 51.87 | 65.37 |
| TPN [13] | 52.78 \pm 0.27 | 66.59 \pm 0.28 |
| EGNN(No Trans) [8] | - | 66.85 |
| R2-D2 [2] | 51.80 \pm 0.20 | 68.4 \pm 0.20 |
| Ours(Conv4) | 51.72 \pm 0.67 | 67.80 \pm 0.30 |
| Ours(Our backbone) | 53.35\pm 0.77 | 69.35\pm 0.61 |

Table 2. Mean accuracies (%) of different methods on the CIFAR-FS dataset. Results are obtained over 600 test episodes with 95% confidence intervals.

| Model | CIFAR-FS (5-way) | |
|------------------------|------------------|----------------|
| | 1-shot | 5-shot |
| MAML [3] | 58.9 \pm 1.9 | 71.5 \pm 1.0 |
| PROTOTYPICAL NETS [19] | 55.5 \pm 0.7 | 72.0 \pm 0.6 |
| RN [20] | 55.0 \pm 1.0 | 69.3 \pm 0.8 |
| GNN [17] | 61.9 | 75.3 |
| R2-D2 [2] | 62.3 \pm 0.2 | 77.4 \pm 0.2 |
| Ours | 61.43 | 76.16 |

Table 3. Ablation study w.r.t. average accuracies (%) over 600 test episodes with 95% confidence intervals MiniImageNet in task 5-way K-shot about ablation study, where $K = 1, 3, 5, 7$ and 10.

| Ave Acc | 5-1 | 5-3 | 5-5 | 5-7 | 5-10 |
|-------------|--------------|--------------|--------------|--------------|--------------|
| RN [20] | 50.44 | 60.63 | 65.32 | 67.73 | 69.81 |
| RPN | 52.43 | 62.96 | 67.03 | 69.51 | 72.01 |
| RMN | 50.54 | 63.12 | 68.28 | 70.49 | 72.12 |
| Ours | 53.35 | 63.94 | 69.35 | 70.87 | 73.17 |



Thanks for watching

<https://github.com/chrisyxue/RMN-RPN-for-FSL>