

Figure 7: Illustration of layer aggregation in GradMix

## A APPENDIX

A.1 SETTINGS FOR OSR EXPERIMENTS

Following section 4.1, the basic descriptions of the six splitting protocols for OSR are in table 6.

Pr	otocols	Known (‡ Classes / Source)	Unknown (‡ Classes / Source)	Openness
Ν	INIST	6 / MNIST	4 / MNIST	22.54%
S	SVHN	6 / SVHN	4 / SVHN	22.54%
Cl	FAR10	6 / CIFAR10	4 / CIFAR10	22.54%
CI	FAR+10	4 / CIFAR10	10 / CIFAR100	46.55%
CI	FAR+50	4 / CIFAR10	50 / CIFAR100	72.784%
Tiny	Imagenet	20 / TinyImagenet	180 / TinyImagenet	68.37%

Table 6: Datasets splitting protocols for known and unknown classes

## A.2 Illustration of Layer Aggregation

737 In 4.1, we tested GradMix for OSR using multiple layers in deep models. The process of layer 738 aggregation is illustrated in figure 7. Assume that the activation maps computed using layer 1, 2, and 739 *N* according to 3.2 are  $M_{layer_1}$ ,  $M_{layer_2}$ , and  $M_{layer_N}$ , the aggregated activation map  $M_{layer_{all}}$  is 740 the summation of them,  $M_{layer_1} = M_{layer_1} + M_{layer_2} + M_{layer_N}$ , after the upsampling of lower 741 resolution activation maps.

A.3 MORE VISUALIZATIONS

Following section 5, the activation maps on more samples are demostrated in figure 8. Same as in 5, GradMix models focus on a broader range of regions in the data.

- A.4 CODE FOR GRADMIX

Our code is publicly released on https://anonymous.4open.science/r/
comprehensive\_osr-8EEF. Part of the frameworks on (self-)supervised contrastive
learning and MoCo are from https://github.com/HobbitLong/SupContrast and
https://github.com/facebookresearch/moco respectively.



trained with and without GradMix. First Row: model trained with GradMix; Second Row: model trained without GradMix