

Fig. 1: Weights generation process. We visualize the weights generation process for two module layers with two modules each.

## **APPENDIX**

## A. Operational Details of the Weight Generation Network

We employ a pre-trained DistilBERT model to encode the natural language task instructions into fixed-length vector in  $\mathbb{R}^{768}$ . The encoded vector, denoted as  $\mathbf{z}_{\text{instr}}$ , is then used as input to a weight generation network, which outputs the weights  $\alpha_1,\ldots,\alpha_k$  for the parameterized modules. This component is used throughout all Meta-World ML-10 and ML-45 training and evaluations.

We now illustrate how the weight generation network operates. For example, consider a setting with two module layers, each containing two modules. The weight generation network computes the weight probabilities  $\mathbf{p}^{l=1}$  between the first and second module layers, and  $\mathbf{p}^{l=2}$  between the second module layer and the output layer. A schematic diagram of this process is provided in Figure 1.

The probabilities  $\mathbf{p}^{l=1} = (p_{11}, p_{12}, p_{21}, p_{22})$  are computed as follows:

$$\mathbf{p}^{l=1} = \operatorname{softmax} \left( \mathbf{W}_d^{l=1} \cdot \operatorname{ReLU}(\mathbf{z}_{\text{instr}}) \right), \tag{1}$$

where  $\mathbf{z}_{\text{instr}}$  is a D-dimensional task instruction vector, and  $W_d^{l=1} \in \mathbb{R}^{4 \times D}$  is a fully connected layer that projects the instruction embedding into a probability vector over all possible connections between the modules in the first and second layers.

The probabilities  $\mathbf{p}^{l=2}=(p_3,p_4)$  are computed as:

$$\mathbf{p}^{l=2} = \operatorname{softmax} \left( \mathbf{W}_d^{l=3} \cdot \operatorname{ReLU} \left( \mathbf{W}_d^{l=2} \mathbf{p}^{l=1} \cdot \mathbf{z}_{\text{instr}} \right) \right), \quad (2)$$

where  $W_d^{l=2} \in \mathbb{R}^{D \times 4}$  maps the vectorized connection probabilities  $p^{l=1}$  back into the task embedding space, and  $W_d^{l=3} \in \mathbb{R}^{2 \times D}$  projects this embedding to produce a probability vector over all possible connections between the modules in the second and output layers.