## 1 Appendix.B Persudo Code of Neural Q-Learning

## $\bf Algorithm~1$ Asynchronous NAF - N collector threads and 1 trainer thread Initialize replay memory D and memory size N, training step M Initialize Q value approximator network weights with random values $\omega$ Initialize the last experience number to train EXPInitialize Training batch size Lfor step = 1, M do \\collect experience action $a_t = \epsilon - \text{greedy}(action list)$ execute $a_t$ and observe state $s_{t+1}$ and reward $r_t$ append $(s_t, a_t, r_t, s_{t+1})$ into memory D $s_t = s_{t+1}$ if step > EXP then randomly sample a batch of transitions $(s_i, a_i, r_i, s_{i+1})$ from replay mem $y_i = r_i + \gamma \max_a' Q(s_{i+1}, a'|\omega)$ train Q value approximator with $y_i$ , $s_i$ pair batch end if end for for t=1,T do Execute $u_t$ and observe $r_t$ and $x_{t+1}$

In our experiments, we set training step M=30000. We build a 5-layer feedforward network to serve as the Q-value approximator, the size of each layer is: 135,270,108,52,8. Gredient descent is used to train the network.

end for