## Training

```
initialize Replay-Memory (capacity N) D
initialize ACTION-VALUE-FUNCTION(random weight \theta) Q
initialize Target-Action-Value-Function(\theta^- = \theta) \hat{Q}
for episode = 1 to M
        initialize State S, Temp-Buffer G
        initialize Collabortor-Probability-Vector(Uniform) p
        for step = 1 to T
                \tilde{r} = \text{discount sum of reward from } S \text{ to } S'
                \tilde{\gamma} = \gamma^{step}
               \tilde{S} = \text{Concatenate}(S, p)
               \tilde{S}' = \text{Concatenate}(S', p)
               store transition (\tilde{S},g,\tilde{r},\tilde{S}') in D
               a_t = \begin{cases} random \ action & \text{with probality } \epsilon \\ \underset{a}{\operatorname{argmax}} \ Q_{\theta}(s, g) & \text{otherwise} \end{cases}
               execute a_t, observe Reward r_t, S = \text{Next-State } S'
                update p with normalized-Behavior-Vector
               sample Minibatch <\tilde{S},g,\tilde{r},\tilde{S}',\tilde{\gamma}> from D
               \begin{split} y &= \tilde{r} + \tilde{\gamma} \cdot \max Q_{\theta} \left( \tilde{S}', \operatorname*{argmax}_{g} Q_{\theta^{-}}(\tilde{S}, g) \right) \\ \mathbf{perform} \ \mathrm{SGD} \ \mathbf{on} \ [y - Q_{\theta}(\tilde{S}, g)]^{2} \ \mathrm{with} \ \mathrm{respect} \ \mathrm{to} \ \theta \end{split}
                reset \hat{Q} = Q every C steps
        resample from p every C' steps
```