Terminal Estimation

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In general cases:

$$Q(s,a) = \sum_{s' \neq \widetilde{s}} p(s'|s,a) V(s') + p(\widetilde{s} \mid s,a) V(\widetilde{s})$$

where \tilde{s} means terminal state so we can set:

$$\begin{split} \hat{Q}(s_t, a_t) &= \sum_{s' \neq \widetilde{s}} p(s'|s, a) V(s') \\ f(s_t, a_t) &= p(\widetilde{s} \mid s, a) V(\widetilde{s}) \\ V(s_t) &= \sum_{a} \pi(a_t \mid s_t) Q(s_t, a_t) \\ Q(s_t, a_t) &= \hat{Q}(s_t, a_t) + f(s_t, a_t) \end{split}$$

In Network, all output is random and doesn't equal zero. for some cases: if p(s'|s,a) = 0, f(s,a) should be zero always and if p(s'|s,a) = 1, $\hat{Q}(s,a)$ should be zero always, too.

so for all cases

$$\hat{Q}(s_t, a_t) = \alpha * \hat{Q}(s_t, a_t) + (1 - \alpha) * (R_{t+1} - \gamma Q(s_{t+1}, a_{t+1}))$$

$$f(s_t, a_t) = \beta * f(s_t, a_t) + (1 - \beta) * \tilde{R}_{t+1}$$

It has a shrinking effect in network and correct the former bias. And it works well in experiment.