## Reinforcement Learning Weekly Written Assignment #2

Author: CS17S011 Ajay Kumar Pandey

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## 1 Solution:

Probability approximately correct(PAC) bound for Multi armed bandit(MAB):

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Quarter elimination algorithm
    Input:\epsilon > 0, \delta > 0
    Output: An optimal arm
    repeat
        sample every arm a \in \delta_l for \frac{2}{\epsilon^2} \log \frac{7}{3\delta_l} times
        find the median of sample means Q_l(a) at round l, denoted by m_l
        s_{l+1} = s_l \setminus \{a : Q_l(a) < m_l\}
        \epsilon_{l+1} = \frac{3}{4}\epsilon_{l}; \delta_{l+1} = \frac{\delta_{l}}{2}; l = l+1
        until|s_l| = 1;
      This algorithm runs log_{4/3}K rounds K:number of arms
At every round there are two bad events, so summing up there bad proba-
bilities should less than \delta_l
event e_1 = \{Q(a_l^*) < q_*(a_l^*) - \epsilon_1/2\};
P[e_1] \leq 3\delta_l/7
In the 2nd bad event we get,
P[\#Badarms \ge 3|s_l|/4 / Q_l(a_l^*) \ge q_*(a_l^*) - \epsilon/2] \le \frac{4\delta_l}{7}
So happening of something bad together is \frac{3\delta_l}{7} + \frac{4\delta_l}{7} = \delta_l
Hence probability of good is 1 - \delta_l
showing sample complexity:
\delta_1 = \delta/2; \delta_l = \delta_{l-1}/2 = \delta/2^l
\epsilon_1 = \epsilon/4; \epsilon_l = (3/4)\epsilon_{l-1} = (3/4)^{l-1}\epsilon/4

k_1 = k; k_l = (3/4)k_{l-1} = (3/4)^{l-1}k
Therefore we have,
\sum_{l=1}^{\log_{4/3}k} k_l \frac{\log(7/3\delta_l)}{\epsilon^2/2} = \sum_{l=1}^{\log_{4/3}k} \frac{2k \log \frac{2^l 7}{\delta 3}}{(4/3)^{l-1} ((3/4)^{l-1} \epsilon/4)^2}
= 32 \sum_{l=1}^{\log_{4/3}k} k(4/3)^{l-1} \left[ \frac{\log(7/3)}{\epsilon^2} + \frac{\log(1/\delta)}{\epsilon^2} + l \frac{\log(2)}{\epsilon^2} \right]
\leq 32k \frac{\log(1/\delta)}{\epsilon^2} \sum_{l=1}^{\infty} (4/3)^{l-1} (lc_1 + c_2)
\leq O(k \frac{\log(1/\delta)}{2})
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Bound for both Median and Quarter elimination algorithm is same.