

## 7 APPENDIX

### 7.1 A Nutshell of Cloud Platform Scoring

#### 7.1.1 Scoring Function

$$s_t = w_1 \cdot m_{ask}(\cdot) + w_2 \cdot m_{dep}(\cdot) + w_3 \cdot m_{uti}(\cdot) + w_4 \cdot m_{af}(\cdot) + w_5 \cdot m_{df}(\cdot) \quad (15)$$

In this function, the user has 3 types of actions: *ask* some quota, *deploy* some quota, and *wait*. Let  $\mathcal{H}^t$  be the user's latest action records, which can be treated as a truncated history with a traceback window  $|\mathcal{H}^t|$ . For each action in  $\mathcal{H}^t$ , it can be denoted as a tuple  $(a_i, q_i)$  where  $a_i$  denotes the action type and  $q_i$  is the corresponding quota. For example,  $a_i = 0$  means this action is "asking  $q_i$  quota";  $a_i = 1$  means "deploying  $q_i$  quota", and  $a_i = 2$  means "waiting for nothing" ( $q_i = 0$ );  $I(\cdot)$  is an indicator;  $\lambda > 0$  measures the degree of time decay. A larger  $\lambda$  means the factor focuses more on recent choices.  $\gamma > 0$  is a scaled parameter.  $q_i^{max}$  is max available quota when action  $a_i$  happens in  $\mathcal{H}^t$ .

To improve the utilization, we wish the user can ask for less quota and deploy the quota as much as possible. Then we can design three factors: ask factor  $m_{ask}$ , deploy factor  $m_{dep}$ , and utilization factor  $m_{uti}$  to encourage users to improve their utilization:

$$m_{ask}(t, \{\gamma_{ask}, \lambda_{ask}\}, \mathcal{H}^t) = \tanh\left(-\gamma_{ask} \cdot \frac{\sum_{i=0}^{|\mathcal{H}^t|} q_i \cdot e^{-\lambda_{ask}(|\mathcal{H}^t|-i-1)} \cdot I(a_i = 0)}{\sum_{i=0}^{|\mathcal{H}^t|} I(a_i = 0)}\right) + 1 \quad (16)$$

$$m_{dep}(t, \{\gamma_{dep}, \lambda_{dep}\}, \mathcal{H}^t) = \tanh\left(\gamma_{dep} \cdot \frac{\sum_{i=0}^{|\mathcal{H}^t|} q_i \cdot e^{-\lambda_{dep}(|\mathcal{H}^t|-i-1)} \cdot I(a_i = 1)}{\sum_{i=0}^{|\mathcal{H}^t|} I(a_i = 1)}\right) \quad (17)$$

$$m_{uti}(t, \{\gamma_{uti}, \lambda_{uti}\}, \mathcal{H}^t) = \tanh\left(\gamma_{uti} \cdot \frac{\sum_{i=0}^{|\mathcal{H}^t|} \frac{q_i}{q_i^{max}} \cdot e^{-\lambda_{uti}(|\mathcal{H}^t|-i-1)}}{\sum_{i=0}^{|\mathcal{H}^t|} e^{-\lambda_{uti}(|\mathcal{H}^t|-i-1)}}\right) \quad (18)$$

To improve users' activeness, we wish the user can take effective actions as many as possible. We design two factors to evaluate the user's activity: ask frequency factor  $m_{af}$ , and deploy frequency factor  $s_{df}$ :

$$m_{af}(t, \{\gamma_{af}, \lambda_{af}\}, \mathcal{H}^t) = \tanh\left(-\gamma_{af} \cdot \frac{\sum_{i=0}^{|\mathcal{H}^t|} I(a_i = 0) \cdot e^{-\lambda_{af}(|\mathcal{H}^t|-i-1)}}{\sum_{i=0}^{|\mathcal{H}^t|} e^{-\lambda_{af}(|\mathcal{H}^t|-i-1)}}\right) + 1 \quad (19)$$

$$m_{df}(t, \{\gamma_{df}, \lambda_{df}\}, \mathcal{H}^t) = \tanh\left(\gamma_{df} \cdot \frac{\sum_{i=0}^{|\mathcal{H}^t|} I(a_i = 1) \cdot e^{-\lambda_{df}(|\mathcal{H}^t|-i-1)}}{\sum_{i=0}^{|\mathcal{H}^t|} e^{-\lambda_{df}(|\mathcal{H}^t|-i-1)}}\right) \quad (20)$$

#### 7.1.2 Reward

The reward of the counter-empirical attacker:

$$r_t = \begin{cases} -0.01 & \text{if } q_t^{dep} > q_{t-1}^{dep}, s_t > s_{t-1} \\ 1.1 * |\Delta s| & \text{if } q_t^{dep} > q_{t-1}^{dep}, s_t \leq s_{t-1} \\ -0.02 & \text{if } q_t^{dep} \leq q_{t-1}^{dep}, s_t < s_{t-1} \\ 1.1 * |\Delta s| & \text{if } q_t^{dep} \leq q_{t-1}^{dep}, s_t \geq s_{t-1} \\ -0.06 & \text{if } q_t^{ask} > 0, s_t < s_{t-1} \\ 1.1 * |\Delta s| & \text{if } q_t^{ask} > 0, s_t \geq s_{t-1} \\ 0 & \text{otherwise} \end{cases} \quad (21)$$

where  $\Delta s = s_t - s_{t-1}$ . Note that each time only one condition occurs in the above function.

### 7.2 A Nutshell of Financial Credit Scoring

#### 7.2.1 Scoring Function

$$s^t = w_1 m_{rep}(\cdot) + w_2 m_{con}(\cdot) + w_3 m_{tra}(\cdot) + w_4 m_{wai}(\cdot) \quad (22)$$

Here, users have four action types: *repay* ( $a_i = 0$ ), *consume* ( $a_i = 1$ ), *in-out* ( $a_i = 2$ ), and *inactivity* ( $a_i = 3$ ). Let  $q^{rep}$  and  $q^{con}$  denote the repaid amount and consumed amount respectively. We consider four factors to encourage good behavior: repay factor  $m_{rep}$ , consume factor  $m_{con}$ , trade-off factor  $m_{tra}$ , wait factor  $m_{wai}$ :

$$m_{rep}(t, \{\gamma_{rep}, \lambda_{rep}\}, \mathcal{H}^t) = \tanh\left(\gamma_{rep} \sum_{i=0}^{|\mathcal{H}^t|} I(a_i = 0) q_i^{rep} \cdot e^{-\lambda_{rep}(|\mathcal{H}^t|-i-1)}\right) \quad (23)$$

$$m_{con}(t, \{\gamma_{con}, \lambda_{con}\}, \mathcal{H}^t) = \tanh\left(-\gamma_{con} \sum_{i=0}^{|\mathcal{H}^t|} I(a_i = 1) q_i^{con} \cdot e^{-\lambda_{con}(|\mathcal{H}^t|-i-1)}\right) + 1 \quad (24)$$

$$m_{tra}(t, \{\gamma_{tra}, \lambda_{tra}\}, \mathcal{H}^t) = \tanh\left(-\gamma_{tra} \sum_{i=0}^{|\mathcal{H}^t|} I(a_i = 2) \left|\frac{q_i^{con}}{q_i^{rep}} - 1\right| \cdot e^{-\lambda_{tra}(|\mathcal{H}^t|-i-1)}\right) + 1 \quad (25)$$

$$m_{wai}(t, \{\gamma_{wai}, \lambda_{wai}\}, \mathcal{H}^t) = \tanh\left(-\gamma_{wai} \sum_{i=0}^{|\mathcal{H}^t|} I(a_i = 3) \cdot e^{-\lambda_{wai}(|\mathcal{H}^t|-i-1)}\right) + 1 \quad (26)$$

#### 7.2.2 Reward

$$r_t = \begin{cases} -q^{rep} * |\Delta s| & \text{if } a_t = 0, s_t > s_{t-1} \\ q^{rep} * |\Delta s| & \text{if } a_t = 0, s_t \leq s_{t-1} \\ q^{con} * |\Delta s| & \text{if } a_t = 1, s_t > s_{t-1} \\ -q^{con} * |\Delta s| & \text{if } a_t = 1, s_t \leq s_{t-1} \\ |q^{con}/q^{rep} - 1| * |\Delta s| & \text{if } a_t = 2, s_t > s_{t-1} \\ -|q^{con}/q^{rep} - 1| * |\Delta s| & \text{if } a_t = 2, s_t \leq s_{t-1} \\ 0 & \text{otherwise} \end{cases} \quad (27)$$