Theorem: When  $\omega_1 > 0$ , there exist an optimal defender strategy that is a pure strategy. Proof:

 $R_i^d$ ,  $P_i^d$ ,  $R_i^a$ ,  $P_i^a$  are the reward (R) and penalty (P) for the defender (superscript d) and the attacker (superscript a) respectively for target *i*. Let  $x = \langle x_i \rangle$  be a defender strategy, and  $x_i$  is the coverage probability of target *i*. Let *q* be the attacking probability for target *i*. According to SUQR model,

$$q_i = \frac{e^{\omega_1 x_i + \omega_2 R_i^a + \omega_3 P_i^a}}{\sum_i e^{\omega_1 x_j + \omega_2 R_j^a + \omega_3 P_j^a}}$$

Let  $U_i^d$  be the defender's expected utility when target *i* is attacked, i.e.,  $U_i^d = (R_i^d - P_i^d)x_i + P_i^d$ . Then defender's overall expected utility can be represented as

$$f(x) = \sum_{i} q_i U_i^d$$

Assume  $\bar{x}$  is the optimal defender strategy. Let  $\bar{S}$  be the set of targets with positive coverage probability, i.e.,  $\bar{S} = \{i | \bar{x}_i > 0\}$ . Then  $\forall i \in \bar{S}, \frac{\partial f}{\partial x_i} | \bar{x} \ge 0$ . Otherwise, a defender strategy with a lower coverage probability on target i will achieve a higher defender expected utility than  $\bar{x}$ , contradict with the optimality. Formally, let  $\Delta_i = (0, 0, ..., \delta, 0, 0)$  be a vector with an infinitesimal positive value in  $i^{th}$  row. If  $\frac{\partial f}{\partial x_i} < 0$ , then  $f(\bar{x} - \Delta_i) = f(\bar{x}) - \delta \frac{\partial f}{\partial x_i} | \bar{x} > f(\bar{x})$ .

Formally, let  $\Delta_i = (0, 0, ..., \delta, 0, 0)$  be a vector with an infinitesimal positive value in  $i^{th}$  row. If  $\frac{\partial f}{\partial x_i} < 0$ , then  $f(\bar{x} - \Delta_i) = f(\bar{x}) - \delta \frac{\partial f}{\partial x_i}|_{\bar{x}} > f(\bar{x})$ . Further, the targets in  $\bar{S}$  can be devided into two subsets  $\bar{S}_1$  and  $\bar{S}_2$  where  $\bar{S}_1 = \{i | \bar{x}_i = 1\}$  and  $\bar{S}_2 = \{i | 0 < \bar{x}_i < 1\}$ . Then  $\forall i, j \in \bar{S}_2, \frac{\partial f}{\partial x_i}|_{\bar{x}} = \frac{\partial f}{\partial x_j}|_{\bar{x}} \ge 0$ . Otherwise, a defender strategy that moves a little bit coverage probability from a target with higher partial derivative to a target with a lower partial derivative will achieve a higher defender expected utility than  $\bar{x}$ , contradict with the optimality. Formally, if  $\frac{\partial f}{\partial x_i}|_{\bar{x}} > \frac{\partial f}{\partial x_j}|_{\bar{x}}$ ,

$$f(\bar{x} + \Delta_i - \Delta_j) - f(\bar{x}) = f(\bar{x} + \Delta_i - \Delta_j) - f(\bar{x} + \Delta_i) + f(\bar{x} + \Delta_i) - f(\bar{x})$$

$$= -\delta \frac{\partial f}{\partial x_j}|_{\bar{x}+\Delta_i} + \delta \frac{\partial f}{\partial x_i}|_{\bar{x}}$$
(2)

$$= -\delta(\frac{\partial f}{\partial x_j}|_{\bar{x}} + \delta \frac{\partial f^2}{\partial x_i \partial x_j}|_{\bar{x}}) + \delta \frac{\partial f}{\partial x_i}|_{\bar{x}}$$
(3)

$$= \delta\left(\frac{\partial f}{\partial x_i}\Big|_{\bar{x}} - \frac{\partial f}{\partial x_j}\Big|_{\bar{x}}\right) - \delta^2 \frac{\partial f^2}{\partial x_i \partial x_j}\Big|_{\bar{x}}$$
(4)

$$> 0$$
 (5)

The last inequality is achieved by neglecting the second order term. So  $f(\bar{x} + \Delta_i - \Delta_j) > f(\bar{x})$ .

Now we show that when  $\omega_1 > 0$ , moving a little a small coverage probability from one target in  $\bar{S}_2$  to another will not decrease the defender's expected utility, i.e.,  $f(\bar{x} + \Delta_i - \Delta_j) - f(\bar{x}) \ge 0$  when  $\omega_1 > 0$ ,  $i, j \in \bar{S}_2$ . Thus, we can always move the coverage probability on targets in  $\bar{S}_2$  until some targets are coverd with probability 1 and others are covered with probability 0. Thus we get a new defender strategy  $\tilde{x}$  with no less expected defender utility than  $\bar{x}$ and the corresponding  $\tilde{S}_2 = \emptyset$ . To prove this, select two targets  $i, j \in \bar{S}_2$ , as  $\frac{\partial f}{\partial x_i}|_{\bar{x}} = \frac{\partial f}{\partial x_j}|_{\bar{x}}, f(\bar{x} + \Delta_i - \Delta_j) - f(\bar{x}) = -\delta^2 \frac{\partial f^2}{\partial x_i \partial x_j}|_{\bar{x}}$  according to line (4). As

$$\frac{\partial f}{\partial x_i}|_{\bar{x}} = \omega_1 q_i (U_i^d - f) + q_i (R_i^d - P_i^d) \ge 0$$

, we have

$$\omega_1(f - U_i^d) \le R_i^d - P_i^d \tag{6}$$

, and similarly,

$$\omega_1(f - U_j^d) \le R_j^d - P_j^d \tag{7}$$

So

$$\frac{\partial f^2}{\partial x_i \partial x_j} = \omega_1^2 q_i q_j (2f - U_i^d - U_j^d) - \omega_1 q_i q_j (R_i^d - P_i^d + R_j^d - P_j^d) \quad (8)$$

$$= \omega_1 q_i q_j (\omega_1 (f - U_i^d) + \omega_1 (f - U_j^d) - (R_i^d - P_i^d + R_j^d - P_j^d)) \quad (9)$$

$$\leq \omega_1 q_i q_j (R_i^d - P_i^d + R_j^d - P_j^d - (R_i^d - P_i^d + R_j^d - P_j^d)) \quad (10)$$

$$= 0 \quad (11)$$

The inequality in line (10) comes from (6) and (7) and the fact the  $\omega_1 > 0$ . Thus we have  $f(\bar{x} + \Delta_i - \Delta_j) - f(\bar{x}) = -\delta^2 \frac{\partial f^2}{\partial x_i \partial x_j}|_{\bar{x}} \ge 0$ . So we can move the coverage probability between targets in  $\bar{S}_2$  without a degredation in defender's expected utility. As a result, we get a new optiaml defender strategy  $\tilde{x}$  whose coverage probabilities are chosen only from 0, 1, i.e., a pure defender strategy.