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By a pass of  $i$  units we mean one which, if complete, has an expected gain of  $i$  distance units. The available distances are  $i = 1, \dots, k$ . Let  $p_i$  denote the probability that such a pass is complete, let  $q_i$  denote the probability that the pass is incomplete, and suppose the probability that it is intercepted is  $\beta q_i$  for some number  $\beta$ . Then from

$$p_i + q_i + \beta q_i = 1$$

it follows that

$$q_i = \frac{1 - p_i}{1 + \beta}. \quad (1)$$

Let  $x_i$  denote the defense resources devoted to stopping a pass of  $i$  units. Then  $p_i$  is a decreasing function of  $x_i$ .

Let  $C_i$  denote the offense's probability of winning if a pass of  $i$  units is complete, and let  $I_i$  be their win probability if the pass is incomplete. Then when the pass is thrown, the offense's probability of winning is

$$f_i(x_i) = p_i(x_i)C_i + \frac{1 - p_i(x_i)}{1 + \beta}I_i. \quad (2)$$

Obviously, the offense will choose an  $i$  for which  $f_i(x_i)$  is maximized. Therefore, the defense has to choose  $x_1, \dots, x_k$  to minimize the maximum of the  $f_i(x_i)$ , subject to the constraint that  $\sum_i x_i \leq X$ , where  $X$  represents the total resources available. We can express the defense's optimization problem as

$$\min_{v, x_1, \dots, x_k} v$$

subject to

$$v \geq f_i(x_i) \quad \text{for all } i, \quad (3)$$

$$\sum_{i=1}^k x_i \leq X, \quad \text{and} \quad (4)$$

$$x_i \geq 0 \quad \text{for all } i. \quad (5)$$

The Lagrangean is

$$v + \sum_{i=1}^k \theta_i [f_i(x_i) - v] + \lambda \left[ \sum_{i=1}^k x_i - X \right] - \mu_1 x_1 - \dots - \mu_k x_k \quad (6)$$

where  $\lambda$ , the  $\theta_i$ , and the  $\mu_i$  are non-negative Lagrange multipliers. The optimality conditions (in addition to the constraints) are

$$1 - \sum_{i=1}^k \theta_i = 0, \quad (7)$$

$$\theta_i f'_i(x_i) + \lambda - \mu_i = 0 \quad \text{for all } i, \quad (8)$$

$$\theta_i [f_i(x_i) - v] = 0 \quad \text{for all } i, \quad (9)$$

$$\lambda \left[ \sum_{i=1}^k x_i - X \right] = 0, \quad \text{and} \quad (10)$$

$$\mu_i x_i = 0 \quad \text{for all } i. \quad (11)$$

The resource constraint 4 will be binding, so  $\lambda$  will be positive.

Now if  $f_i(x_i) \neq v$  for some  $i$ , then  $\theta_i = 0$  by equation 9. Then, by equation 8,  $\mu_i$  cannot be zero, and hence  $x_i = 0$  by equation 11. So for every  $i$ , either  $f_i(x_i) = v$ , or  $x_i = 0$ . In words, the optimal strategy for the defense is to use its resources in such a way that every pass distance gives the same win probability to the offense—except for those distances that the defense ignores entirely. Let  $\bar{v}$  and  $\bar{x}_1, \dots, \bar{x}_k$  denote the solution to the defense's optimization problem. There is actually an easy way to solve for these values: Given any  $v$ , let  $\bar{x}_i(v)$  be the solution to  $v = f_i(x_i)$ , if there exists a positive  $x_i$  satisfying that equation, and let  $\bar{x}_i(v) = 0$  otherwise. Then  $\bar{v}$  is the value of  $v$  for which  $\sum_i \bar{x}_i(v) = X$ , and the  $\bar{x}_i(\bar{v})$  are the defense's optimal resource allocation.

In equilibrium, the optimal strategy for the offense will, in general, be a randomized strategy, in which a pass of  $i$  units will be chosen with probability  $\pi_i$ . In the model it's unnecessary to know the offense's randomization probabilities; we require only  $\bar{v}$ . However, we can easily determine the  $\pi_i$  by the condition that the  $\bar{x}_i$  be optimal against them. Given the  $\pi_i$ , the defense's optimization problem can be alternatively expressed as

$$\min_{x_1, \dots, x_k} \sum_{i=1}^k \pi_i f_i(x_i)$$

subject to  $\sum_i x_i - X \leq 0$ , and  $-x_i \leq 0$  for all  $i$ . The Lagrangean is

$$\sum_{i=1}^k \pi_i f_i(x_i) + \lambda \left[ \sum_{i=1}^k x_i - X \right] - \mu_1 x_1 - \dots - \mu_k x_k \quad (12)$$

where  $\lambda$  and the  $\mu_i$  are non-negative Lagrange multipliers. We know that the  $\bar{x}_i$  are the solution to this optimization problem. Therefore the  $\bar{x}_i$  satisfy the optimality conditions, which (in addition to the constraints) are

$$\pi_i f'_i(\bar{x}_i) + \lambda - \mu_i = 0 \quad \text{for all } i, \quad (13)$$

$$\lambda \cdot \left( \sum_{i=1}^k \bar{x}_i - X \right) = 0, \quad \text{and} \quad (14)$$

$$\mu_i \bar{x}_i = 0 \quad \text{for all } i. \quad (15)$$

Equations 13, 14, and 15 can now be regarded as equations for the  $\pi_i$ . If  $\bar{x}_i = 0$ , we can simply set  $\pi_i = 0$ . Otherwise,  $\mu_i = 0$ , and equation 13 together with the requirement that  $\sum_i \pi_i = 1$  determines the  $\pi_i$ .