Defense Efficiency

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Measure of Efficiency: Why?

While defending against attacks, we need to

- evaluate the defenses.
- learn (ML)
- etc.

To this end, we need to define efficiency of defense.



It is important that the efficiency will

- reflect the cost-benefit tradeoff
- be higher when the system recovers

Approach



- formally model
- 2 look for a useful sets of properties
- **3** suggest a natural satisfying solution
- 4 add properties to characterize the solution
- 5 Generalize the solution

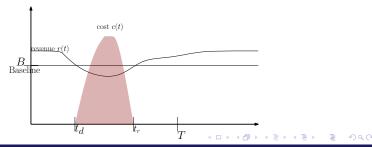
Model

- **1** Revenue $r : \mathbb{R} \to \mathbb{R}_+$
- 2 Time bound T
- **③** Detection and recovery times t_d and t_r relatively to B

4 Impact
$$I \stackrel{\Delta}{=} \int_{t_d}^{t_r \text{ or } T} (B - r(t)) dt$$

5 Cost $c \colon \mathbb{R} \to \mathbb{R}$

6 Total cost
$$Ct \stackrel{\Delta}{=} \int_{t_d}^{t_r \text{ or } T} (c(t)) dt$$



- Decreasing with impact I
- Decreasing with total cost Ct
- No recovery is always smaller than recovery
- E: {recovered, not recovered} $\times \mathbb{R}_+ \times \mathbb{R}_+ \rightarrow [0, 1]$

Definition

Define E(recovered or not, I, Ct) as

$$\begin{cases} \beta + \alpha \frac{B \cdot T - I}{B \cdot T} + (1 - \beta - \alpha) \frac{C \cdot T - Ct}{C \cdot T} \\ = 1 - \frac{\alpha}{B \cdot T} I - \frac{1 - \beta - \alpha}{C \cdot T} Ct & Recovered, \\ \alpha(\frac{\beta}{1 - \beta}) \frac{B \cdot T - I}{B \cdot T} + (1 - \beta - \alpha)(\frac{\beta}{1 - \beta}) \frac{C \cdot T - Ct}{C \cdot T} \\ = \beta - \alpha \frac{\beta}{(1 - \beta)(B \cdot T)} I - (1 - \beta - \alpha) \frac{\beta}{(1 - \beta)(C \cdot T)} Ct & otherwise. \end{cases}$$
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Axiomatic Characterization Theorem

$$\begin{cases} \beta + \alpha \frac{B \cdot T - I}{B \cdot T} + (1 - \beta - \alpha) \frac{C \cdot T - Ct}{C \cdot T} \\ = 1 - \frac{\alpha}{B \cdot T} I - \frac{1 - \beta - \alpha}{C \cdot T} Ct \\ \alpha(\frac{\beta}{1 - \beta}) \frac{B \cdot T - I}{B \cdot T} + (1 - \beta - \alpha)(\frac{\beta}{1 - \beta}) \frac{C \cdot T - Ct}{C \cdot T} \\ = \beta - \alpha \frac{\beta}{(1 - \beta)(B \cdot T)} I - (1 - \beta - \alpha) \frac{\beta}{(1 - \beta)(C \cdot T)} Ct \quad \text{otherwise.} \end{cases}$$

is the unique definition that satisfies the following:

- Linearly decreasing with I
- Linearly decreasing with Ct
- The ratio of the linear coefficient of the impact to the linear coefficient of the total cost is independent of the recovery
- If no recovery takes place, exactly all the values in [0, β] are obtained; otherwise, exactly the values in [β, 1] are obtained

Generalization to More Inputs



- More variables (e.g., multiple revenues and costs)
- Each variable has positive or negative influence
- Want to arrive at a measure between 0 and 1

Required Properties

Let f be a strictly increasing function

- Increasing with each $f(y_i), i = 1, ..., m$
- Decreasing with each $f(x_j), j = m + 1, \dots, m + l$
- No recovery is always smaller than recovery
- E: {recovered, not recovered} $imes \mathbb{R}^{m+l}_+ o [0,1]$



Let nonnegative $\alpha_1, \alpha_2, \ldots, \alpha_{m+l-1}$ s.t. $\sum_{i=1}^{m+l-1} \alpha_i \leq 1 - \beta$ define the importance of each term. Let each y_i be bounded by Y_i and let each x_j be bounded by X_j .

Definition

Define $E(recovered or not, y_1, \ldots, y_m, x_{m+1}, \ldots, x_{m+l})$ as

$$\begin{cases} \beta + \sum_{i=1}^{m} \alpha_i \frac{f(y_i)}{f(Y_i)} + \sum_{j=m+1}^{m+l-1} \alpha_j \frac{f(X_j) - f(x_j)}{f(X_j)} \\ + (1 - \beta - \sum_{k=1}^{m+l-1} \alpha_k) \frac{f(X_{m+l}) - f(x_{m+l})}{f(X_{m+l})} \\ \sum_{i=1}^{m} \alpha_i (\frac{\beta}{1-\beta}) \frac{f(y_i)}{f(Y_i)} + \sum_{j=m+1}^{m+l-1} \alpha_j (\frac{\beta}{1-\beta}) \frac{f(X_j) - f(x_j)}{f(X_j)} \\ + (1 - \beta - \sum_{k=1}^{m+l-1} \alpha_k) (\frac{\beta}{1-\beta}) \frac{f(X_{m+l}) - f(x_{m+l})}{f(X_{m+l})} \quad otherwise. \end{cases}$$
(2)

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$$\sum_{i=1}^{m} \alpha_{i} \left(\frac{\beta}{1-\beta}\right) \frac{f(y_{i})}{f(Y_{i})} + \sum_{j=m+1}^{m+l-1} \alpha_{j} \left(\frac{\beta}{1-\beta}\right) \frac{f(X_{j}) - f(x_{j})}{f(X_{j})} \\ + \left(1 - \beta - \sum_{k=1}^{m+l-1} \alpha_{k}\right) \left(\frac{\beta}{1-\beta}\right) \frac{f(X_{m+l}) - f(x_{m+l})}{f(X_{m+l})} & otherwise. \end{cases}$$

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Definition

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$$\begin{cases} \beta + \sum_{i=1}^{m} \alpha_{i} \frac{f(y_{i})}{f(Y_{i})} + \sum_{j=m+1}^{m+l-1} \alpha_{j} \frac{f(X_{j}) - f(x_{j})}{f(X_{j})} \\ + (1 - \beta - \sum_{k=1}^{m+l-1} \alpha_{k}) \frac{f(X_{m+l}) - f(x_{m+l})}{f(X_{m+l})} & Recovered, \end{cases}$$

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Expanding Equation: Axiomatic Characterization Theorem

$$\begin{cases} \beta + \sum_{i=1}^{m} \alpha_i \frac{f(y_i)}{f(Y_i)} + \sum_{j=m+1}^{m+l-1} \alpha_j \frac{f(X_j) - f(x_j)}{f(X_j)} \\ + (1 - \beta - \sum_{k=1}^{m+l-1} \alpha_k) \frac{f(X_{m+l}) - f(x_{m+l})}{f(X_{m+l})} \\ \sum_{i=1}^{m} \alpha_i (\frac{\beta}{1-\beta}) \frac{f(y_i)}{f(Y_i)} + \sum_{j=m+1}^{m+l-1} \alpha_j (\frac{\beta}{1-\beta}) \frac{f(X_j) - f(x_j)}{f(X_j)} \\ + (1 - \beta - \sum_{k=1}^{m+l-1} \alpha_k) (\frac{\beta}{1-\beta}) \frac{f(X_{m+l}) - f(x_{m+l})}{f(X_{m+l})} \end{cases}$$
Recovered,

is the unique definition that satisfies the following:

- Linearly increasing with each $f(y_i)$
- 2 Linearly decreasing with each $f(x_j)$
- The ratio of the linear coefficient of f(y_i) or f(x_j) to the linear coefficient of any other f(y_k) or f(x_p) is independent of the recovery
- If no recovery takes place, exactly the values in [0, β] can be obtained; otherwise, all the values in [β, 1] and only they can be obtained

Generalization: Combining Efficiency Values

Definition

Given E_i , the *i*th component of the efficiency, and its weight γ_i , define

$$E \stackrel{\Delta}{=} \sum_{i=1}^{n} \gamma_i E_i, \tag{3}$$

where
$$\gamma_i \geq 0$$
, $\sum_{i=1}^n \gamma_i = 1$.

Each $E_i \in [0,1] \Rightarrow E \in [0,1]$.



The two generalizations are not equivalent. However,

Theorem

If the system recovers, they are equivalent!

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1 Defined efficiency that is monotone and respects recovery

2 Axiomatic characterization

Generalizations:

- Expanding equation and characterizing
- Combining efficiency values as black boxes

Comparing these generalizations

