

Plans Are Worthless but Planning Is Everything: A Theoretical Explanation of Eisenhower's Observation

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Eisenhower's Observation

Rational Decision ...

In These Terms, What ...

Options

Estimating L_0

Estimating L_1

Conclusions

Acknowledgments

Home Page

Title Page

⏪

⏩

◀

▶

Page 1 of 14

Go Back

Full Screen

Close

Quit

1. Eisenhower's Observation

- Dwight D. Eisenhower was:
 - the Supreme Commander of the Allied Expeditionary Forces in Europe during WW2
 - and later the US President.
- He emphasized that his war experience taught him that “plans are worthless, but planning is everything”.
- At first glance, this sounds paradoxical: if plans are worthless, why bother with planning at all?
- In this paper, we show that this Eisenhower's observation has a meaning:
 - while following the original plan in constantly changing circumstances is often not a good idea,
 - the existence of a pre-computed original plan enables us to produce an almost-optimal strategy.

2. Rational Decision Making: a Brief Reminder

- According to decision making theory:
 - decisions by a rational decision maker
 - can be described as maximize the value a certain function known as utility.
- E.g., in financial situations, when a company needs to make a decision, the overall profit can be used as utility.
- To describe a possible action x , we usually need to describe the values of several quantities x_1, \dots, x_n .
- E.g., a decision about a plant involves selecting amounts x_i of manufactured gadgets of different type.
- Similarly, we need several quantities a_1, \dots, a_m to describe a situation.
- Let $u(x, a)$ denote the utility that results from performing action x in situation a .

3. In These Terms, What Is Planning

- Let \tilde{a} describe the original situation.
- Based on this situation, we come up with an action \tilde{x} that maximizes the corresponding utility:

$$u(\tilde{x}, \tilde{a}) = \max_x u(x, \tilde{a}).$$

- Computing this optimal action \tilde{x} is what we usually call *planning*.
- When we need to start acting, the situation may have changed to $a \neq \tilde{a}$.
- Let us denote the corresponding change by $\Delta a \stackrel{\text{def}}{=} a - \tilde{a}$, then $a = \tilde{a} + \Delta a$.

4. Options

- One possibility is to simply ignore the change, and apply the original plan \tilde{x} to the new situation $a = \tilde{a} + \Delta a$.
- This plan is, in general, not optimal for the new situation.

- The actually optimal plan is x^{opt} for which

$$u(x^{\text{opt}}, \tilde{a} + \Delta a) = \max_x u(x, \tilde{a} + \Delta a).$$

- In comparison with the optimal plan, we lose the amount $L_0 \stackrel{\text{def}}{=} u(x^{\text{opt}}, \tilde{a} + \Delta a) - u(\tilde{x}, \tilde{a} + \Delta a)$.
- Why cannot we just find the optimal solution for the new situation?
- Optimization is NP-hard, so, it is not possible to find the exact optimum in reasonable time.

5. Options (cont-d)

- What we can do is:
 - try to use some feasible algorithm – e.g., solving a system of linear equations,
 - to modify the plan \tilde{x} into $\tilde{x} + \Delta x$.
- Due to NP-hardness, this feasibly modified plan is, in general, not optimal.
- We hope that the resulting loss $L_1 \stackrel{\text{def}}{=} u(x^{\text{opt}}, \tilde{a} + \Delta a) - u(\tilde{x} + \Delta x, \tilde{a} + \Delta a)$ is much smaller than L_0 .
- In this paper, we show that indeed $L_1 \ll L_0$; so:
 - even if L_0 is so large that the original plan is worthless,
 - the modified plan may leads to a reasonably small loss $L_1 \ll L_0$.
- This explains Eisenhower's observation.

6. Estimating L_0

- We assume that the difference Δa is reasonably small.
- So, the corresponding difference in action $\Delta x^{\text{opt}} \stackrel{\text{def}}{=} x^{\text{opt}} - \tilde{x}$ is also small.
- We can therefore expand L_0 in Taylor series and keep only terms linear and quadratic in Δx :

$$L_0 = u(x^{\text{opt}}, \tilde{a} + \Delta a) - u(x^{\text{opt}} - \Delta x^{\text{opt}}, \tilde{a} + \Delta a) =$$

$$\sum_{i=1}^n \frac{\partial u}{\partial x_i}(x^{\text{opt}}, \tilde{a} + \Delta a) \cdot \Delta x_i^{\text{opt}} +$$

$$\frac{1}{2} \cdot \sum_{i=1}^n \sum_{i'=1}^n \frac{\partial^2 u}{\partial x_i \partial x_{i'}}(x^{\text{opt}}, \tilde{a} + \Delta a) \cdot \Delta x_i^{\text{opt}} \cdot \Delta x_{i'}^{\text{opt}} + o((\Delta a)^2).$$

- By definition, the action x^{opt} maximizes $u(x, \tilde{a} + \Delta a)$.
- Thus, we have $\frac{\partial u}{\partial x_i}(x^{\text{opt}}, \tilde{a} + \Delta a) = 0$.

7. Estimating L_0 (cont-d)

- So, the above expression for L_0 takes the simplified form

$$L_0 = \frac{1}{2} \cdot \sum_{i=1}^n \sum_{i'=1}^n \frac{\partial^2 u}{\partial x_i \partial x_{i'}} (x^{\text{opt}}, \tilde{a} + \Delta a) \cdot \Delta x_i^{\text{opt}} \cdot \Delta x_{i'}^{\text{opt}} + o((\Delta a)^2).$$

- Δx_i^{opt} can be estimated from the condition:

$$\frac{\partial u}{\partial x_i} (x^{\text{opt}}, \tilde{a} + \Delta a) = \frac{\partial u}{\partial x_i} (\tilde{x} + \Delta x^{\text{opt}}, \tilde{a} + \Delta) = 0.$$

- For $a = \tilde{a}$, u is max when $x = \tilde{x}$, so $\frac{\partial u}{\partial x_i} (\tilde{x}, \tilde{a}) = 0$.
- Expanding the equation in Taylor series in Δx_i and Δa_j and taking $\frac{\partial u}{\partial x_i} (\tilde{x}, \tilde{a}) = 0$ into account, we get:

$$\sum_{i'=1}^n \frac{\partial^2 u}{\partial x_i \partial x_{i'}} (\tilde{x}, \tilde{a}) \cdot \Delta x_{i'}^{\text{opt}} + \sum_{j=1}^m \frac{\partial^2 u}{\partial x_i \partial a_j} (\tilde{x}, \tilde{a}) \cdot \Delta a_j + o(\Delta x, \Delta a) = 0.$$

[Home Page](#)[Title Page](#)[◀◀](#) [▶▶](#)[◀](#) [▶](#)

Page 9 of 14

[Go Back](#)[Full Screen](#)[Close](#)[Quit](#)

8. Estimating L_0 (final)

- Thus, the first approximation Δx_i to the values Δx_i^{opt} satisfies a system of linear equations:

$$\sum_{i'=1}^n \frac{\partial^2 u}{\partial x_i \partial x_{i'}}(\tilde{x}, \tilde{a}) \cdot \Delta x_j = - \sum_{j=1}^m \frac{\partial^2 u}{\partial x_i \partial a_j}(\tilde{x}, \tilde{a}) \cdot \Delta a_j.$$

- A solution to a system of linear equations is a linear combination of the right-hand sides.
- Thus, the values Δx_i are a linear function of Δa_j .
- Substituting these linear expressions into the formula for L_0 , we conclude that L_0 is quadratic in Δa_j :

$$L_0 = \sum_{j=1}^m \sum_{j'=1}^m k_{jj'} \cdot \Delta a_j \cdot \Delta a_{j'} + o((\Delta a)^2) \text{ for some } k_{jj'}.$$

9. Estimating L_1

- The 1st approximation Δx to the difference Δx^{opt} can be obtained by solving a system of linear equations.
- How much do we lose if we use $x^{\text{lin}} = \tilde{x} + \Delta x$?
- Here, $\Delta x^{\text{opt}} = \Delta x + \delta x$, where δx is of 2nd order in Δx and Δa : $\delta x = O((\Delta a)^2)$.
- The loss L_1 of using $x^{\text{lin}} = x^{\text{opt}} - \delta x$ instead of x^{opt} is:

$$L_1 = u(x^{\text{opt}}, \tilde{a} + \Delta a) - u(x^{\text{lin}}, \tilde{a} + \Delta a) = \\ u(x^{\text{opt}}, \tilde{a} + \Delta a) - u(x^{\text{opt}} - \delta x, \tilde{a} + \Delta a).$$

- If we expand this expression in δx , we get:

$$L_1 = \sum_{i=1}^n \frac{\partial u}{\partial x_i}(x^{\text{opt}}, \tilde{a} + \Delta a) \cdot \delta x_i +$$

$$\frac{1}{2} \cdot \sum_{i=1}^n \sum_{i'=1}^n \frac{\partial^2 u}{\partial x_i \partial x_{i'}}(x^{\text{opt}}, \tilde{a} + \Delta a) \cdot \delta x_i \cdot \delta x_{i'} + o((\delta x)^2).$$

10. Estimating L_1 (cont-d)

- Since x^{opt} is the action that, for $a = \tilde{a} + \Delta a$, maximizes utility, we get $\frac{\partial u}{\partial x_i}(x^{\text{opt}}, \tilde{a} + \Delta a) = 0$.
- Thus, the expression for L_1 gets a simplified form

$$L_1 = \frac{1}{2} \cdot \sum_{i=1}^n \sum_{i'=1}^n \frac{\partial^2 u}{\partial x_i \partial x_{i'}}(x^{\text{opt}}, \tilde{a} + \Delta a) \cdot \delta x_i \cdot \delta x_{i'} + o((\delta x)^2).$$

- We know that the values δx_i are quadratic in Δa .
- Thus, we conclude that for the modified action, *the loss L_1 is a 4-th order function of Δa_j* :

$$L_1 = \sum_{j=1}^m \sum_{j'=1}^m \sum_{j''=1}^m \sum_{j'''=1}^m k_{jj'j''j'''} \cdot \Delta a_j \cdot \Delta a_{j'} \cdot \Delta a_{j''} \cdot \Delta a_{j'''} + o((\Delta a)^5).$$

11. Conclusions

- We conclude that:
 - the loss L_0 related to using the original plan is quadratic in Δa , while
 - the loss L_1 related to using a feasibly modified plan is of 4th order in terms of Δa .
- For small Δa , we have $L_1 \sim (\Delta a)^4 \ll L_0 \sim (\Delta a)^2$.
- Let $\varepsilon > 0$ be the maximum loss that we tolerate.
- Since $L_1 \ll L_0$, we have three possible cases:
 - (1) $\varepsilon < L_1$, (2) $L_1 \leq \varepsilon \leq L_0$, and (3) $L_0 < \varepsilon$.
- In the 1st case, even the modified action does not help.
- In the 3rd case, the change in the situation is so small that it is Ok to use the original plan \tilde{x} .

12. Conclusions (cont-d)

- In the second case, we have exactly the Eisenhower situation:
 - if we use the original plan \tilde{x} , the resulting loss L_0 much larger than we can tolerate;
 - in this sense, the original plan is worthless;
 - on the other hand, if we feasible modify the original plan into x^{lin} , then we get an acceptable action.
- So, we indeed get a theoretical justification of Eisenhower's observation.

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Rational Decision ...

In These Terms, What ...

Options

Estimating L_0

Estimating L_1

Conclusions

Acknowledgments

Home Page

Title Page

◀◀

▶▶

◀

▶

Page 13 of 14

Go Back

Full Screen

Close

Quit

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Eisenhower's Observation

Rational Decision . . .

In These Terms, What . . .

Options

Estimating L_0

Estimating L_1

Conclusions

Acknowledgments

Home Page

Title Page



Page 14 of 14

Go Back

Full Screen

Close

Quit