

Lecture 11: The Valuation and Pricing of Information

I. Economic Foundations of Value of Information

		Coin state ω		Prior belief $q(\omega)$
		1/2 tail	1/2 head	
action	0	2t	0	
	1	0	2t	

$u(a, \omega)$

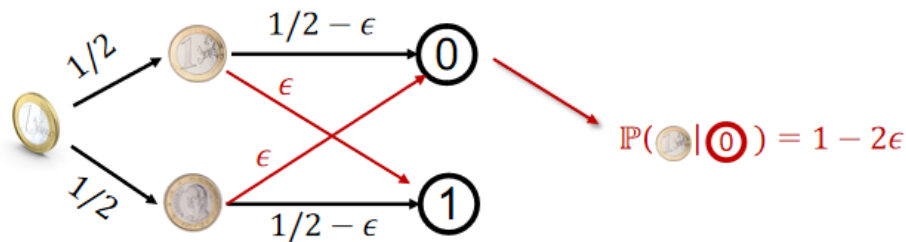
Without information, decision maker(DM) gets $\max_a [\mathbb{E}_{\omega \sim q} u(a, \omega)] = t$

With my (full) information, DM gets $\mathbb{E}_{\omega \sim q} [\max_a u(a, \omega)] = 2t$

Value of (full) info = $\mathbb{E}_{\omega \sim q} [\max_a u(a, \omega)] - \max_a [\mathbb{E}_{\omega \sim q} u(a, \omega)]$

		1 - 2ε	2ε	Posterior belief $p(\omega)$, given ①
		1/2 tail	1/2 head	
action	0	t	0	Prior belief $q(\omega)$
	1	0	3t	

$u(a, \omega)$



Noisy information revelation

Realistically, can think of 0 as noisy prediction of state ω (e.g., stock trend, purchase prob)

Question: What is the value of this noisy signal 0?

- Without knowing this signal, DM takes action 1
- With this signal 0, DM takes action 0 (assuming ε very small)
- However, true distribution is the posterior p regardless

$$\text{Value of knowing } 0 = \mathbb{E}_{\omega \sim p}[u(0, \omega)] - \mathbb{E}_{\omega \sim p} u(1, \omega)$$

Definition(FK'19): Consider an arbitrary decision making problem $u(a, \omega)$, suppose a signal updates the DM's belief about state ω from $q \in \Delta(\Omega)$ to $p \in \Delta(\Omega)$, the value of this signal is defined as

$$D^u(p; q) = \mathbb{E}_{\omega \sim p}[u(a^*(p), \omega)] - \mathbb{E}_{\omega \sim p} u(a^*(q), \omega)$$

Example 1:

- $a \in A = \Delta(\Omega) \rightarrow$ action is to pick a distribution over states
- $u(a, \omega) = \log a_\omega$
- Which action $a \in \Delta(\Omega)$ maximizes expected utility $\mathbb{E}_{\omega \sim p}[u(a, \omega)]$?

$$a^*(p) = p$$

$$D^u(p; q) = \sum_{\omega} p_{\omega} \log \frac{p_{\omega}}{q_{\omega}} \quad \text{KL-divergence}$$

Example 2:

- $a \in A = \Delta(A) \rightarrow$ action is to pick a distribution over states
- $u(a, \omega) = -\|a - e_{\omega}\|^2$

$$a^*(p) = p$$

$$D^u(p; q) = \|p - q\|^2 \quad \text{Squared distance}$$

Some obvious properties

- Non-negativity: $D^u(p; q) \geq 0$
- Null information has no value: $D^u(q; q) = 0$
- Order-invariant: if DM receives signal σ_1, σ_2 , the order of receiving them does not affect final expected total value

In this case, we say $D(p; q)$ is a valid measure for value of information

Theorem 1(FK'19): Consider any $D(p; q)$ function. There exists a decision problem $u(a, \omega)$ such that

$$D^u(p; q) = \mathbb{E}_{\omega \sim p}[u(a^*(p), \omega)] - \mathbb{E}_{\omega \sim p} u(a^*(q), \omega)$$

if and only if $D(p; q)$ satisfies

- Non-negativity: $D^u(p; q) \geq 0$
- Null information has no value: $D^u(q; q) = 0$
- Order-invariant: if DM receives signal σ_1, σ_2 , the order of receiving them does not affect final expected total value

Theorem 2(FK'19):

1. For any concave function H , its Bregman divergence is a valid measure for value of information.
2. Conversely, for any valid measure $D(p; q)$ for value of information,

$$H(q) = \sum_{\omega} q^{\omega} D(e_{\omega}, q)$$

is a concave function whose Bregman divergence is $D(p; q)$

Why useful?

- Many functions - even natural ones like l_2 distance $\|p - q\|$ - are not valid measures

- In fact, any metric is not valid, since metric cannot be a Bregman divergence
- There are efficient ways to tell whether a $D(p, q)$ is valid



II. Optimal Pricing of Information



III. Summary and Open Problems