

# **Expected-Utility Preference Reversals in Information Acquisition**

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## **Abstract**

Suppose you must choose between two pieces of information A and B. In the absence of cost, you would prefer to obtain A rather than B, and in fact would be willing to take more risk to obtain A than B. Nevertheless, you would pay more money for B than for A. Are your preferences consistent with expected utility? The answer is yes, they may very well be. We give an example to illustrate how this may happen, and relate this reversal phenomenon to the well-known discrepancy between buying and selling prices for lotteries. The existence of such reversals dispels any notion that the relative value of competing information acquisitions should not depend on the nature of the acquisition. Among expected utility maximizers, only those with constant risk attitude avoid this phenomenon.

# 1 Introduction

Suppose you must choose between acquiring information about A and acquiring information about B. In the absence of cost, you would prefer to learn about A rather than B, and in fact would be willing to take more risk to learn about A. Nevertheless, you would pay more money to learn about B than A. Are your preferences consistent with expected utility? The answer is yes, they may very well be. A closely related question involves certainty equivalents (cash equivalents, selling prices) and buying prices for lotteries. As is well known, buying and selling prices need not coincide unless risk attitude is constant. However, consider an expected utility maximizer who would pay more for lottery  $X$  than lottery  $Y$ . Might she demand a higher selling price for  $Y$  if she owned it than she would for  $X$  if she owned it? Once again the answer is yes.

In what follows, we give examples to illustrate how these reversals may happen. We also show that among expected utility maximizers, only those with constant risk attitude avoid these phenomena, either for lottery payoffs or for information. While the discrepancies between buying and selling prices of lotteries are well known, the question of whether preference reversals are possible and which utility functions allow them seems not to have been treated adequately in the literature. Bell (1988) establishes the type of utility functions that allow only one preference reversal over the entire range of wealth levels for the case of lotteries, without reference to buying or selling. Our emphasis here is on buying and selling information which involves changes in decision strategy as well.

In Section 2 we begin by briefly reviewing the literature on information value, and discussing measures of information value. Our approach involves reducing questions about valuing information to the analogous questions about valuing payoff lotteries, and this approach is presented in Section 3, where we present our main results. As the reader will see, the relation between buying information and buying a lottery is more delicate than one might first guess. In Section 4 we provide some concluding remarks.

## 2 Information Value

### Background

Early work in the area of information value as it pertains to decision analysis is attributable to Howard (1966, 1967) and Matheson (1968). Their consideration of the value of clairvoyance led to the concept of perfect information and a methodology for calculating the expected value of perfect information. General discussions of information value may be found in Raiffa (1968), Gould (1974), and Howard (1988). Rothkopf (1971) advocates EVPI as a measure of venture risk. Hazen and Felli (1997) propose information value on input parameters as the proper way to measure problem sensitivity. In recently developed normative expert systems (e.g., Heckerman 1991), information value is used to determine what question to next ask the user.

Information value is notorious for its lack of convenient mathematical properties. For example, it is well known that information value is not additive across sources (see Howard 1988 for a discussion). LaValle (1968), Gould (1974), and Hilton (1981) show the lack of any general relationship between information value and the level of wealth, the degree of absolute or relative risk aversion, or the Rothschild-Stiglitz degree of uncertainty in the prior. Miller (1975) examines situations where it is possible to obtain information sequentially throughout the decision process and determines that the value of any particular piece of information is a function of the prices of all other obtainable pieces of information. In a production model with uncertain demand, Merkofer (1977) shows that the value a decision maker places on a given piece of information depends on the flexibility of his decisions. In the non-expected utility framework, it is well known that nonlinearity in probability can give rise to negative value of information (e.g., Wakker 1988). More details about information evaluation under non-additive expected utility theory can be found in LaValle and Xu (1990).

### Measures of information value

Suppose a decision maker will receive uncertain payoff  $V_a$  when choosing action  $a$ . Suppose  $V_a$  depends directly or indirectly on an uncertainty  $X$ . We assume the decision maker acts to maximize expected utility under some utility function  $u$  defined, continuous

and increasing over payoffs. Let  $a^*$  be an optimal action in the absence of further information, the event which we denote  $I_{\mathcal{E}}$ . Letting  $V$  be the overall payoff, we have

$$E[u(V) | I_{\emptyset}] = \max_a E[u(V_a)] = E[u(V_{a^*})].$$

Let  $I_X$  denote the event that the value of the uncertain quantity  $X$  will be available prior to choosing, and let  $a^*(x)$  be an action maximizing  $E[u(V_a) | X = x]$ . Then

$$E[u(V) | I_X] = E_X \left[ \max_a E[u(V_a) | X] \right] = E_X \left[ E[u(V_{a^*(X)}) | X] \right] = E[u(V_{a^*(X)})].$$

The standard approach to quantifying information value is to ask what the decision maker would give up to acquire the information. The *buying price*  $BPI_X$  is defined as the maximum payoff the decision maker would forgo to learn  $X$  before choosing. It satisfies:

$$E[u(V - BPI_X) | I_X] = E[u(V) | I_{\emptyset}].$$

If we let  $CE[V | I_X] = u^{-1}(E[u(V) | I_X])$  be the *certainty equivalent* of  $V$  given  $I_X$ , then we may write the last equality as

$$CE[(V - BPI_X) | I_X] = CE[V | I_{\mathcal{E}}].$$

The *utility increase*  $EUI_X$  is defined as the increase in utility obtained by being able to observe  $X$  before choosing:

$$EUI_X = E[u(V) | I_X] - E[u(V) | I_{\mathcal{E}}].$$

$EUI$  is more mathematically tractable than  $BPI$ , and has therefore seen some use in theoretical contexts (e.g., Bernardo and Smith 1994). A related measure of information value is the *certainty equivalent increase*  $CEI$ , defined as

$$CEI_X = CE[V | I_X] - CE[V | I_{\mathcal{E}}].$$

Another measure is the *selling price* of an information. The selling price of  $I_X$ , denoted by  $SPI_X$ , is the minimum price a seller who already possesses  $I_X$  (but has not used it) would ask for giving up  $I_X$ , and it satisfies

$$E[u(V) | I_X] = E[u(V + SPI_X) | I_{\emptyset}].$$

Yet another measure of information value is the *Probability Price* which is the maximum probability of a large, designated loss that the decision maker is willing to bear in order to acquire the information. To define it formally, suppose there is a payoff  $v_0$  which bounds below the payoff variables  $V_a$  in the decision problem at hand. Imagine that in exchange for  $I_X$ , the decision maker can take on some chance  $p$  of obtaining  $v_0$  instead of  $V_a$ . The probability price of  $I_X$ , denoted by  $PPI_X$ , is the probability that satisfies

$$PPI_X u(v_0) + (1 - PPI_X) E[u(V)|I_X] = E[u(V)|I_{\emptyset}].$$

Now let us consider the question of how different measures rank order different information. For any  $I_X$  the three measures, namely,  $EUI_X$ ,  $CEI_X$  and  $SPI_X$  are readily seen to be increasing transformations of one another. Coming to  $PPI_X$ , we can rewrite its defining equation as

$$\frac{PPI_X}{1 - PPI_X} = \frac{EUI_X}{E[u(V)|I_{\emptyset}] - u(v_0)}$$

revealing that it is also an increasing transformation of  $EUI_X$  and therefore of  $CEI_X$  and  $SPI_X$ . Thus all four measures will always rank different information in the same order.

Let us say that two measures of information value are *ordinally equivalent* if they rank information values identically. We then have the following formal assertion.

PROPOSITION 1: *CEI*, *EUI*, *SPI* and *PPI* are ordinally equivalent measures of information value, that is, for any two uncertainties  $X$  and  $W$ ,

$$CEI_X \geq CEI_W \Leftrightarrow EUI_X \geq EUI_W \Leftrightarrow SPI_X \geq SPI_W \Leftrightarrow PPI_X \geq PPI_W.$$

In what follows we will use any one or another of these four measures for comparison with the fifth measure, *BPI*. The reason we keep track of all four measures is that each one has its own advantage. *CEI* is relevant for the delta property discussed below, *EUI* is tractable, *SPI* has a practical significance, and *PPI* has the flavor of buying the information.

When two measures of information value are not ordinally equivalent, there will exist reversals, pairs  $X$ ,  $W$  of uncertainties such that the information value of  $X$  is higher than that of  $W$  according to one measure but lower according to the other. The primary

question we address in this paper is whether the measure  $BPI$  is ordinally equivalent to  $SPI$  (and therefore to  $CEI$ ,  $EUI$  and  $PPI$ ), or in other words, whether there exist what we term  $BP$ - $SP$  reversals for information.

The more specific question of whether  $CEI$  and  $BPI$  are *equal* (not merely ordinally equivalent) can be addressed by invoking familiar results on buying and selling prices. If the utility function  $u$  has constant risk attitude (*i.e.*,  $u$  is linear or exponential), then certainty equivalents obey the *delta property* (Howard 1967)

$$CE[V+D] = CE[V] + D.$$

Invoking the delta property on the defining equation for  $BPI$  yields  $BPI_x = CEI_x$ . This is an extension of the familiar result for lotteries that certainty equivalent and buying price are equal if (and only if) utility is linear or exponential (Raiffa 1968, Howard 1970).

At first glance it may appear that there is no oddity if  $SPI$  and  $BPI$  are not ordinally equivalent. But if they are not, then  $PPI$  and  $BPI$  will also not be ordinally equivalent. This implies that one might rationally reverse the preference for information depending upon how one pays for the information, paying money or accepting risk. Accepting risk to acquire information is not as rare as one might think. For example, medical patients accept the mortality risks of exploratory surgery and the cancer hazard of X-ray examination; and police detectives interrogate dangerous criminals to pursue investigative leads.

In view of the above it is tempting to hope that  $SPI$  and  $BPI$  are ordinally equivalent. In the next section we show that in fact this equivalence fails in general, that is,  $BP$ - $SP$  reversals can happen for information sources. Such reversals are basically the analogies to  $BP$ - $SP$  reversals for simple payoff lotteries.

### 3 Valuing Information and valuing payoff lotteries

#### Examples of reversals

We first illustrate a *BP-SP* reversal for payoff lotteries. Suppose the utility function  $u$  is given by  $u(w) = w^{1/2}$  for nonnegative  $w$ . Let the initial wealth  $L$  be \$100 and the payoff of the lotteries  $Y_1$  and  $Y_2$  be as shown below:



Because  $E[u(L + Y_1)] = 15$  and  $E[u(L + Y_2)] = 15.11$ , the decision maker will prefer receiving  $Y_2$  as a gift. Since  $E[u(L + Y_1)] = 15 = u(\$225)$ , it follows that the certainty equivalent of  $L+Y_1$  is \$225, an increase of \$125 over the initial wealth \$100. Similarly, the certainty equivalent of  $L+Y_2$  is \$228.31, and increase of \$128.31 over initial wealth.

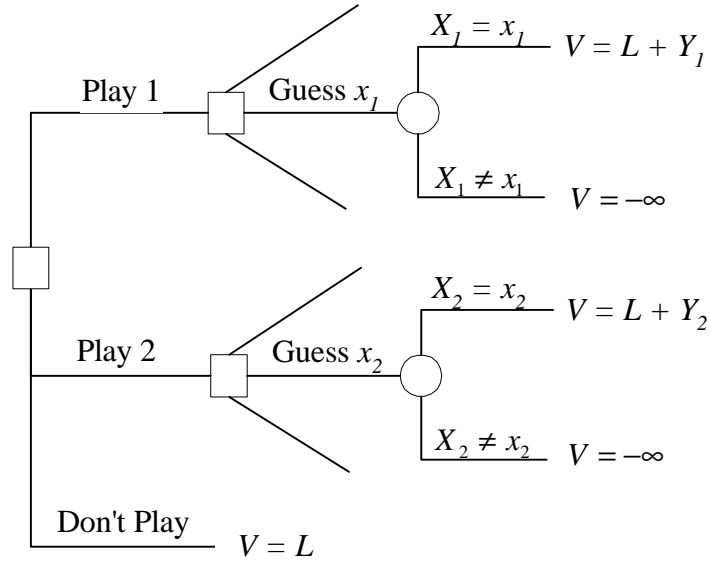
The buying price  $b_i$  for  $Y_i$  satisfies the equation

$$E[u(L + Y_i - b_i)] = u(L)$$

Calculation shows that  $b_1 = \$93.75$  and  $b_2 = \$83.79$ , in contrast to the respective certainty equivalent increases \$125 and \$ 128.31. These respective quantities would be equal if risk attitude were constant.

An analogous reversal can be generated for information acquisition by appropriately embedding  $Y_1$  and  $Y_2$  into a decision problem. Extend the utility function  $u(w) = w^{1/2}$  to negative  $w$  in any way such that  $u$  is increasing, continuous and has  $u(-\infty) = -\infty$ . Consider the decision problem in Figure 1.

**Figure 1: Embedding the payoff lotteries  $Y_1, Y_2$  into a decision problem**



In this problem, the decision maker must choose to play game 1 or game 2 or not to play. If game  $i$  is chosen, then the decision maker must guess the value of a discrete uncertainty  $X_i$ . A correct guess increases wealth by  $Y_i$ , whereas an incorrect guess yields an extremely large penalty (represented by  $-\infty$ ).

It is not difficult to show that the certainty equivalent increases and buying prices for information about the independent uncertainties  $X_1$  and  $X_2$  are identical to the corresponding quantities for the payoff lotteries  $Y_1$  and  $Y_2$  above, that is,

$$\begin{aligned}
 CEI_{X_1} &= \$125 & CEI_{X_2} &= \$128.31 \\
 BPI_{X_1} &= \$93.75 & BPI_{X_2} &= \$83.79.
 \end{aligned}$$

We have therefore exhibited a *BP-CE* and therefore a *BP-SP* reversal for information sources.

### **Payoff lotteries induced by information acquisition choices**

There is, of course, an intimate relationship between the operators *CE* and *BP* over random variables and the operators *CEI* and *BPI* over information sources. Given a payoff lottery (random variable)  $Z$ , let  $CE[Z] = u^{-1}(E[u(Z)])$  denote the certainty

equivalent of  $Z$ , and let  $BP[Z/Wealth=L]$  be the buying price of  $Z$  given (possibly uncertain) wealth  $L$ . If  $BP[Z/Wealth = L] = b$ , then  $b$  satisfies the equation

$$E[u(L+Z - b)] = E[u(L)].$$

Because  $E[u(V)|I_X] = E[u(V_{a^*(X)})]$ , we have

$$\begin{aligned} CEI_X &= u^{-1}(E[u(V)|I_X]) - u^{-1}(E[u(V)|I_{\emptyset}]) \\ &= u^{-1}(E[u(V_{a^*(X)})]) - u^{-1}(E[u(V_{a^*})]) \\ &= CE[V_{a^*(X)}] - CE[V_{a^*}] \end{aligned} \quad (1)$$

In other words,  $CEI_X$  is the difference in certainty equivalents between  $V_{a^*(X)}$  and  $V_{a^*}$ .

The relationship between the buying price operators is more delicate. The buying price  $BPI_X$  is the quantity  $b$  satisfying

$$E[u(V - b) | I_X] = E[u(V) | I_{\emptyset}]$$

or equivalently

$$E_X \left[ \max_a E[u(V_a - b) | X] \right] = E[u(V_{a^*})]$$

Let  $a_{V-b}^*(X)$  be the maximizer on the left side of this equation. The act  $a_{V-b}^*(X)$  is optimal after paying  $b$  for  $I_X$  and observing  $X$ . When risk attitude is not constant,  $a_{V-b}^*(X)$  need not equal  $a^*(X) = a_V^*(X)$ , which is the optimal act after observing  $X$  for free. When  $a_{V-BPI_X}^*(X) = a^*(X)$ , then paying the buying price for  $I_X$  does not affect the optimal act when  $X$  is observed. In this case we say that  $a^*(X)$  is *purchase invariant*. We call the random quantity  $V_{a^*(X)} - V_{a^*}$  the *net information acquisition payoff*. If  $a^*(X)$  is purchase invariant, then the net information acquisition payoff is the same regardless of whether  $I_X$  is purchased or obtained for free. When  $a^*(X)$  is purchase invariant, the last displayed equation becomes

$$E[u(V_{a^*(X)} - BPI_X)] = E[u(V_{a^*})].$$

We have established the following result.

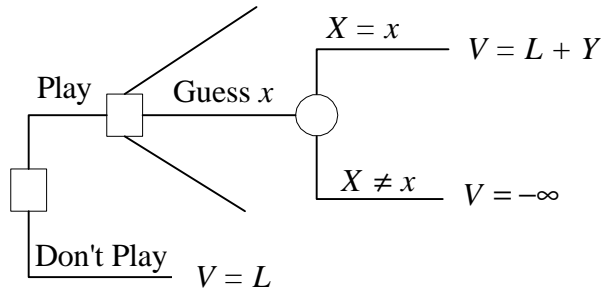
PROPOSITION 2: In a given decision problem with uncertainty  $X$ , suppose  $a^*(X)$  is purchase invariant. Then

$$BPI_X = BP[V_{a^*(X)} - V_{a^*} / \text{Wealth} = V_{a^*}]$$

that is,  $BPI_X$  is the buying price of the net information acquisition payoff  $V_{a^*(X)} - V_{a^*}$  at wealth  $V_{a^*}$ .

Because we wish to relate the properties of the information-value operators  $BPI$ ,  $EUI$ ,  $CEI$  to the buying price and certainty equivalent operators over payoff lotteries, it would be useful in light of Proposition 2 if for any payoff lottery  $Y$ , we could find a decision problem containing an uncertainty  $X$  with  $a^*(X)$  purchase invariant such that the net information acquisition payoff  $V_{a^*(X)} - V_{a^*}$  was  $Y$ . A construction much like the one in Figure 1 is possible, and is depicted in Figure 2.

**Figure 2: A decision problem with net information acquisition payoff equal to the payoff lottery  $Y$ .**



In this problem,  $L$  is the current wealth level, and  $Y$  is a nonnegative payoff lottery. As before, the decision maker must first choose to play or not, and if s/he chooses to play, s/he must guess the value of the discrete uncertainty  $X$ . A correct guess yields independent incremental payoff  $Y$ , but an incorrect guess yields payoff  $-\infty$ . Without further information, it is not optimal to play, so  $a^* = \text{"Don't Play"}$ , and  $V_{a^*} = L$ . But if  $X = x$  can be observed prior to deciding, then because  $Y \geq 0$  and  $u$  is increasing, it is optimal to play and guess  $x$ . Then  $a^*(x) = \text{"Play and guess } x\text{"}$  and  $V_{a^*(X)} = L + Y$ .

If we let  $BP(Y / \text{Wealth} = L) = b$ , then  $b \geq 0$  because  $Y \geq 0$ . Then by definition of  $BP$ , we have

$$E[u(L+Y-b)] = u(L) \stackrel{3}{\geq} u(L-b).$$

The inequality  $E[u(L+Y-b)] \stackrel{3}{\geq} u(L-b)$  implies  $a_{v-b}^*(X) = \text{“Play and guess X”} = a^*(X)$ , and the equality  $E[u(L+Y-b)] = u(L)$  implies that  $b = BPI_X$ . Therefore  $a^*(X)$  is purchase invariant. Moreover  $V_{a^*(X)} - V_{a^*} = L + Y - L = Y$ . Therefore by Proposition 2,  $BPI_X = BP[Y/Wealth = L]$ . Finally it is easy to check that  $CEI_X = CE[Y+L] - CE[L]$ . We have shown the following result.

PROPOSITION 3: For any increasing continuous utility function  $u$ , any wealth level  $L$  and any nonnegative random variable  $Y$ , there is a decision problem with an uncertainty  $X$  such that  $BPI_X = BP[Y/Wealth = L]$  and  $CEI_X = CE[Y+L] - CE[L]$ .

### Equality of certainty equivalent and buying price

Proposition 3 allows us to extend known properties of certainty equivalents and buying prices for payoff lotteries to information sources. The following well known result on payoff lotteries receives general mention with partial or no proof by Howard (1970) and by Raiffa (1968). Because we did not locate a specific proof of necessity in the literature, we provide our own.

PROPOSITION 4: In order that  $BP[Z/Wealth = L] = CE[Z+L] - CE[L]$  for all nonnegative payoff lotteries  $Z$  it is necessary and sufficient that the utility function  $u(x)$  has constant risk attitude (*i.e.*, is linear or exponential).

PROOF: Sufficiency is shown by LaValle (1968) and Howard (1970). To demonstrate necessity, note that buying price has the so-called **D**-property:

$$BP[Z+\mathbf{D} / Wealth = L] = BP[Z/Wealth = L] + \mathbf{D} \quad \mathbf{D} \geq 0.$$

Therefore  $CE[Z+L]$  also possesses the **D**-property, *i.e.*  $CE[Z+L+\mathbf{D}] = CE[Z+L] + \mathbf{D}$  for all  $\mathbf{D} \geq 0$ . Then the risk premium, as defined in Pratt(1964), is constant in wealth, and it follows from results there that  $u$  must have constant risk attitude. ♦

COROLLARY:  $BPI_X = CEI_X$  for all decision problem uncertainties  $X$  if and only if the utility function  $u(x)$  has constant risk attitude (*i.e.*, is linear or exponential).

### Reversals for certainty equivalents and buying prices

Although the question of *BP-CE equality* is treated in the literature, we are aware of no examination of *BP-CE reversals* for payoff lotteries. Say that a utility function  $u$  allows no *BP-CE reversals* if for every pair of payoff lotteries  $X, Y$

$$BP[X] \preceq BP[Y] \hat{U} CE[X] \preceq CE[Y].$$

Here we omit reference to wealth, for expositional simplicity. Of course, linear or exponential utility functions allow no reversals because for them,  $BP = CE$ . Are there any other utility functions that do not allow reversals? We answer that question as follows.

PROPOSITION 5: A utility function  $u$  allows no *BP-SP* reversals for payoff lotteries if and only if its risk attitude is constant (*i.e.*,  $u$  is linear or exponential).

PROOF: Only necessity remains to be established. Switching SP to CE, the assumption of no *BP-SP* reversals implies no *BP-CE* reversals. By simple logical manipulation then

$$CE[X] = CE[Y] \hat{U} BP[X] = BP[Y]$$

From this it follows that *CE* is a *function* of *BP*, that is, there is some function  $g$  such that  $CE[X] = g(BP[X])$  for all  $X$ . But for *constant* payoff lotteries  $X$ ,  $CE[X] = BP[X]$ , so it must be that  $g$  is the identity function. Therefore,  $CE[X] = BP[X]$  for all payoff lotteries  $X$ . ♦

In an analogous way, we may inquire what utility functions allow no *BP-SP* reversals for information sources, that is, for which utility functions *BPI* and *SPI* are ordinally equivalent measures of information value. A nearly identical line of proof leads to the same conclusion, namely, that only linear or exponential utility functions qualify. The only additional point that needs to be established in the proof is that for any  $u$ , there is a

decision problem having an uncertainty  $X$  such that  $CEI_X = BPI_X$ . To specify such a problem, take  $Y$  to be a positive constant in the decision problem of Figure 2. We have established the following result.

PROPOSITION 6:  $BPI$  and  $SPI$  are ordinally equivalent measures of information value for a given utility function  $u$  if and only if  $u$  has constant risk attitude (*i.e.*,  $u$  is linear or exponential).

## 4 Conclusion

We have shown that although utility increase and certainty equivalent increase are ordinally equivalent measures of information value, neither is, in general, ordinally equivalent to buying price, that is, certainty equivalent increase and buying price may rank information sources differently. Only utility functions with constant risk attitude avoid this phenomenon.

A rough explanation of  $BP$ - $SP$  reversals for payoff lotteries is easily devised: A decision maker may prefer  $X$  to  $Y$  at the current wealth level. But payment of a buying price alters wealth level, and when risk attitude is not constant,  $X$  may no longer be preferred to  $Y$  at the new wealth level. This explanation applies as well to information sources, but somehow in this arena,  $BP$ - $SP$  reversals seem more surprising. We suspect that for many, information is a more fundamental entity, in that the relative values of competing information acquisitions should not depend on the nature of the acquisition. This viewpoint must be abandoned in light of the fact that an expected utility maximizer can be willing to take more risk to learn about  $X$  (*i.e.*,  $X$  has a greater utility increase), but pay more to learn about  $Y$  (*i.e.*,  $Y$  has a higher buying price).

The point is not a trivial one:  $EUI$  is often preferred over  $BPI$  as a measure of information value due to its analytical tractability. Neither measure is more correct than the other – they are merely different measures. The possibility of  $BP$ - $SP$  reversal is something that an analyst should be aware of in those circumstances.

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