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Newcomb's Problem

John Collins

Columbia University, New York NY, U.S.A.

Newcomb's problem is a decision puzzle whose difficulty and interest stem from the fact that the possible outcomes are probabilistically dependent on, yet causally independent of, the agent's options. The problem is named for its inventor, the physicist William Newcomb, but first appeared in print in a 1969 paper by Robert Nozick [12]. Closely related to, though less well-known than, the Prisoners' Dilemma, it has been the subject of intense debate in the philosophical literature. After three decades, the issues remain unresolved. Newcomb's problem is of genuine importance because it poses a challenge to the theoretical adequacy of orthodox Bayesian decision theory. It has led both to the development of *causal decision theory* and to efforts aimed at defending the adequacy of the orthodox theory.

This article surveys the debate. The problem is stated in Section 1. Arguments for each of the opposed solutions are rehearsed in Section 2. Section 3 contains a sketch of causal decision theory. One strategy for defending orthodox

Bayesianism is outlined in Section 4.

1 The Problem

There are two boxes on the table in front of you. One of them is transparent and can be seen to contain one thousand dollars. The other is opaque. You know that it contains either one million dollars or nothing. You must decide whether to take (1) only the contents of the opaque box (call this the *one-box option*); or, (2) the contents of both boxes (the *two-box option*). You know that a remarkably accurate Predictor of human deliberation placed the million dollars in the opaque box yesterday if and only if it then predicted that you would choose today to take only the contents of that box. You have great confidence in the Predictor's reliability. What should you do?

2 Opposed Arguments

There is a straightforward and plausible expected value argument for the rationality of the one-box choice. If you take only the contents of the opaque box, then, almost certainly, the Predictor will have predicted this and placed the million dollars in the box. If you choose to take the contents of both boxes, then, almost certainly, the Predictor will have predicted this and left the opaque box empty. Thus the expected value of the one-box choice is very close to one million dollars, while that of the two-box choice is approximately one thousand dollars. The one-box choice *maximizes your expected value*. You

should therefore choose to take only one box.

But there is also a straightforward and plausible dominance argument supporting the two-box choice. The prediction was made yesterday. The million dollars is either there in the opaque box or it is not. Nothing that you do now can change the situation. If the million dollars is there, then the two-box choice will yield one million plus one thousand dollars, while the one-box choice yields only the million. And if the opaque box is empty, there is no million dollars to be had: the two-box choice gives you a thousand, the one-box option leaves you with no money at all. Whichever situation now obtains, you will be one thousand dollars better off if you take both boxes. The two-box choice is your *dominant* option. You should therefore choose to take both boxes.

Care must be taken in applying the second argument, for dominance reasoning is appropriate only when probability of outcome is independent of choice. Example: Nasty Nephew has insulted Rich Auntie who threatens to cut Nephew from her will. Nephew would like to receive the inheritance, but, other things being equal, would rather not apologize. Either Auntie cuts Nephew from the will or she doesn't. If she does, Nephew receives nothing, and in this case prefers not having apologized. If she doesn't, then Nephew receives the inheritance, but once again, prefers inheritance-with-no-apology to inheritance-plus-apology. Does a dominance argument give Nephew reason to refrain from apologizing? Clearly that line of thought is fallacious. It ignores the fact that whether or not Nephew is cut from the will may depend upon whether or not

he apologizes.

Does this observation vitiate the two-boxer's appeal to dominance? That depends on what one means by "dependence". What the Predictor did yesterday is certainly *probabilistically dependent* on what you choose today. The conditional probability that the opaque box contains a million dollars given that you choose the one-box option is close to one, while the conditional probability of the million dollars being present given that you choose to take both boxes is close to zero. But yesterday's prediction is *causally independent* of today's choice, for what is past is now fixed and determined and beyond your power to influence. We are used to these two senses of dependence coinciding, but Newcomb's problem illustrates the possibility of divergence when, for example, there is reliable information to be had about the future. Armed with the distinction, two-boxers maintain that dominance reasoning is valid whenever outcomes are *causally* independent of choice.

The dominance argument can be made even more vivid by considering the following variation on the original problem. The opaque box has a transparent back. Your dearest friend is sitting behind the boxes and can see the contents of both. Your friend is allowed to advise you what you should do. What choice will your friend tell you to make? Obviously: to take both boxes. Now it would be irrational to ignore the advice of a friend who has your best interests at heart and who you know is better informed than you are. And shouldn't the fact that you know *in advance* what your friend would tell you to do in this

variation of the problem determine your opinion about what to do in the original version?

One-boxers reply that the kind of agent who follows the friend's advice problem, and two-boxers in general, will very likely finish up with only a thousand dollars. "If you're so smart," they taunt, "why ain'cha rich?" Two-boxers may counter that Newcomb's problem is simply a situation in which one can expect irrationality to be rewarded. If "choosing rationally" means no more than choosing so as to maximize expected reward, it may seem hard to see how such a response can be coherent, but see [5] and [11].

3 Causal Decision Theory

Those two-boxers who are skeptical that orthodox Bayesian decision theory of the kind developed in [6] can deliver the right answer to Newcomb's problem face the challenge of providing an account of rational decision making consistent with this judgment. By the early 1980's several independently formulated but essentially equivalent versions of an alternative account had appeared. (See, e.g. [2], [5], [8], [10], [13], and [15].) According to this alternative account, now widely referred to as *causal decision theory* (CDT), calculations of expected value should be sensitive to an agent's judgments about the probable causal consequences of the available options. The differences between this causal theory and orthodox Bayesian decision theory are perhaps best appreciated by proceeding from a definition of expected value that is neutral between

the two accounts.

The notion of expected value relates the value assigned by an agent to an option A to the values the agent assigns to the more particular ways w in which A might turn out to be true. All parties to the debate agree that the expected value $V(A)$ of an option A is a probability weighted average of the values of these various w . That is:

$$V(A) = \sum_w V(w) \cdot P_A(w)$$

where P_A is the probability function that results when the agent's subjective probability function P is revised so as to accept A . Where the theories differ is on the method of probability revision deemed appropriate to this context. According to the orthodox Bayesian theory, the appropriate method is revision by *conditionalization* on the option A :

$$V(A) = \sum_w V(w) \cdot P(w/A)$$

whereas causal decision theorists maintain that for each w the weight should be given by the probability of the subjunctive conditional: *if A were chosen then w would be true*. This conditional is usually abbreviated $A \square \rightarrow w$. Thus:

$$V(A) = \sum_w V(w) \cdot P(A \square \rightarrow w)$$

That these two definitions of expected value do not coincide is far from obvious. But in fact they *don't* coincide because they *cannot*. David Lewis, one of

the co-founders of CDT, demonstrated the impossibility of defining a propositional connective $\square\rightarrow$ with the property that for each A, C the probability of the conditional $A\square\rightarrow C$ equals the conditional probability of C given A . (See [9] and the articles on *Counterfactual reasoning, quantitative and qualitative (philosophical aspects)* in this volume).

But the divergence of these two definitions is best seen by applying them to Newcomb's problem. There are two options: the one-box choice (B_1) and the two-box and two-box choice (B_2); and four possible outcomes: receiving nothing (N), receiving one thousand dollars (K), receiving one million dollars (M), or receiving one million one thousand dollars (i.e. "getting the lot" L). Assume that the values you assign to these outcomes are given by the dollar amounts, and suppose finally that you have 99% confidence in the Predictor's reliability, i.e. that you assign the conditional probabilities: $P(M/B_1) = 0.99$, $P(N/B_1) = 0.01$, $P(L/B_2) = 0.01$, $P(K/B_2) = 0.99$.

Then on the orthodox account:

$$V(B_1) = V(M).P(M/B_1) + V(N).P(N/B_1) = 990,000$$

$$V(B_2) = V(L).P(L/B_2) + V(K).P(K/B_2) = 11,000$$

and so according to the Bayesian theory the one-box choice maximizes expected value.

Before the causal theory can be applied, probabilities must be assigned to the

relevant subjunctive conditionals: $B_1 \square \rightarrow M$, $B_1 \square \rightarrow N$, $B_2 \square \rightarrow L$, and $B_2 \square \rightarrow K$. The first of these, $B_1 \square \rightarrow M$, is the conditional: ‘If you were to take only the contents of the opaque box, then you would receive a million dollars.’ Now this conditional is true if and only if the Predictor placed the million dollars in the opaque box yesterday. Similarly the conditional $B_2 \square \rightarrow L$: ‘If you were to take the contents of boxes, then you would receive one million one thousand dollars.’ is true just in case the Predictor put the million in the opaque box yesterday. It follows that:

$$P(B_1 \square \rightarrow M) = P(B_2 \square \rightarrow L) = \mu$$

and

$$P(B_1 \square \rightarrow N) = P(B_2 \square \rightarrow K) = 1 - \mu$$

where μ is whatever probability you assign to the proposition that yesterday the Predictor placed one million dollars in the opaque box. Thus:

$$\begin{aligned} V(B_1) &= V(M).P(B_1 \square \rightarrow M) + V(N).P(B_1 \square \rightarrow N) \\ &= 1,000,000.\mu \end{aligned}$$

$$\begin{aligned} V(B_2) &= V(L).P(B_2 \square \rightarrow L) + V(K).P(B_2 \square \rightarrow K) \\ &= 1,000,000.\mu + 1,000 \end{aligned}$$

and so according to the causal theory the two-box choice maximizes expected value. Note that these assignments reflect the causal decision theorist’s advocacy of the dominance argument in a situation in which outcomes are causally

independent of choice, for to say that C is causally independent of A is just to say that $P(A \square \rightarrow C) = P(C)$.

There are a couple of delicate issues involved here. First of all, one must resist the temptation to say such things as:

“The million dollars is there in the opaque box, but if you were to choose to take only the contents of that box, then the million dollars would not be there.”

a thought to which one might be led by the consideration that:

“If you were to make the one-box choice now, then the Predictor, being so reliable, would have to have predicted yesterday that that was what you would do today.”

Of course these conditionals would be true if your choice today had the power to affect the past, if, for example, your choosing to take only one box could somehow cause the Predictor yesterday to have predicted that that was what you would do. But that’s not so. Newcomb’s problem is not a story about backward causation. The Predictor predicted what you will do, but that prediction was not caused by what you will do.

But of course the two conditionals above are natural enough things to say. The point is not that this sort of *backtracking interpretation* of the subjunctive conditional is impossible—conditionals are sometimes used in precisely this

way. It is rather that backtracking conditionals are not relevant to the present discussion, for they fail to reflect causal structure. So let the causal decision theorist simply stipulate that the choice-to-outcome conditionals not be given a backtracking interpretation. The truth-value of $A \square \rightarrow C$ is evaluated, roughly speaking, by supposing the antecedent A to be true, and then determining whether the consequent C follows from that supposition. “No backtracking” means that in supposing A to be true, one continues to hold true what is past, fixed, and determined. (For further discussion of these issues see the articles on *Counterfactual reasoning* in this volume).

Secondly, it is reasonable to be suspicious of CDT’s appeal to μ —your present unconditional probability that the million dollars is in the opaque box. Given that you know the Predictor to be highly reliable, the best evidence you now have as to whether or not the million dollars is in the box is provided by your current beliefs about what you will choose to do. So μ will have a value close to 1 if you believe that you will take only one box, and a value close to 0 if you think that you will take both boxes. The value of μ thus helps to determine your expected values, while at the same time being determined in part by those expected values. Some opponents of CDT (e.g. Isaac Levi) have found this problematic, maintaining that agents cannot coherently assign probabilities to the outcomes of their own current deliberations, while one advocate of CDT (Brian Skyrms) exploits this very feature of the theory in giving a *dynamic* account of deliberation as a kind of positive feedback mechanism on the agent’s

probabilities that terminates only as the agent's degree of belief that a certain choice will be made approaches the value 1 (see [14]).

Further criticism of CDT has focussed on the *partition problem*. As Levi has pointed out [8], the deliverances of CDT depend upon the particular way in which outcomes are specified. CDT may prescribe different courses of action when one and the same choice situation is described in two different ways. So unless there is some preferred way of partitioning outcomes, CDT is inconsistent. Example: if the outcomes in Newcomb's problem are taken to be (1) you receive the contents of the one (opaque) box only, and (2) you receive the contents of the two boxes, then CDT can be made to prescribe the one-box choice, for these two outcomes are clearly under your, the agent's, causal control.

Defenders of CDT take the moral of this to be that outcomes to a well-posed decision problem must be specified to a degree of detail that is "complete in the sense that any further specification of detail is irrelevant to the agent's concerns" [5]. Some opponents of CDT consider this unnecessarily limiting and a mark of the orthodox theory's superiority, for the evidential theory may be applied in a way that is robust under various different ways of partitioning outcomes (see e.g. the articles by Levi and Eells in [1]).

CDT may be recast in a different form. As mentioned above, Lewis proved in [9] that the probability of a conditional is not, in general, a conditional probability. It is impossible to define a propositional connective $\square\rightarrow$ with the

property that $P(A \square \rightarrow C) = P(C/A)$. However, in that same paper he also showed that if conditionalization is replaced by a kind of probability revision method called *imaging* a similar equation can be maintained.

Revising P by conditionalization on A amounts to removing any probability assigned to not- A and renormalizing. Roughly speaking, revising P by imaging on A amounts to reassigning the prior probability of each way A might be false to the “closest” way to that in which A would be true. Write P_A for the result of revising P to accept A . The imaging revision methods are precisely those which are *linear* in the sense that:

$$\text{If } P = \alpha.P' + (1 - \alpha).P'' \text{ then } P_A = \alpha.(P'_A) + (1 - \alpha).(P''_A)$$

This allows a recharacterization of CDT which never mentions causation. The essential feature of CDT is that its definition of expected value employs a probability revision method that is linear.

(See [3]. For a fuller discussion of imaging, the reader is again referred to the articles on *Counterfactual reasoning* in this volume. The most complete and sophisticated account of CDT to date is to be found in [7].)

4 The Tickle Defense

The relation between Newcomb’s problem and the Prisoners’ Dilemma has been noted and debated (see e.g. [1]). But whereas the Prisoners’ Dilemma

has been widely discussed by social and behavioral scientists in several fields, the discussion of Newcomb's problem has mainly been confined to the philosophical literature. This may be because the significance of the problem is more purely theoretical than practical, but is perhaps due also the somewhat fantastic nature of the Newcomb problem itself (in Richard Jeffrey's phrase: "a prisoners' dilemma for space cadets" [6]).

Less fantastic problems of the Newcomb type may arise, however, in situations in which a statistical correlation is due to a common cause rather than to a direct causal link between the correlated factors. Thus, for example, though it is well established that there is a strong statistical correlation between smoking and lung cancer, it doesn't follow that smoking is a cause of cancer. The statistical association might be due to a common cause (a certain genetic predisposition say) of which smoking and lung cancer are independent probable effects. (This is sometimes referred to as "Fisher's smoking hypothesis"). If you have the bad gene you are more likely to get lung cancer than if you don't, and you are also more likely to find that you prefer smoking to abstaining. Suppose you are convinced of the truth of this common cause hypothesis and like to smoke. Does the fact that smoking increases the probability of lung cancer give you a reason not to smoke?

Surely it doesn't. Smoking is, for you, the dominant option, and it would be absurd for you to choose not to smoke in order to avoid lung cancer if your getting lung cancer is causally independent of your smoking. Here CDT

delivers what is, pretty clearly, the correct answer. On the face of it this would seem to provide a much more clear cut counterexample to orthodox Bayesian theory than Newcomb's problem.

But this is not clearly a counterexample at all. Here the strategy of the defenders of orthodoxy has been to argue that in this more realistic kind of problem, the theory, correctly applied, prescribes the same choice as CDT. Let S be the proposition that you smoke, and let C be the proposition that you get lung cancer. Although you know that smoking and lung cancer are statistically correlated, it doesn't follow that C and S are probabilistically dependent by the lights of your subjective probability assignment. That is because you will typically have further information that will *screen off* the probabilistic dependence of C on S . In the present example, this might consist of your knowing that you have a craving to smoke. Let T be the proposition that you have this craving or "tickle" (the strategy being discussed is customarily referred to as the *tickle defense*). Then $P(C/T \& S) = P(C/T \& \text{not} S)$. Now since you know that you have the tickle, $P(T) = 1$ and hence according to your probability assignment $P(C/S) = P(C/\text{not} S)$. The non-causal probabilistic dependence of C on S has vanished (see [4] but note also the reply by Jackson and Pargetter in [1]).

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