The Strategic Role of the Weight of Evidence in Abduction

Lok C. Chan

1 Introduction

In Progress

2 Peirce on Balance and Weight

While the idea of evidential weight is closely associated with Keynes' discussion in A Treatise, Peirce was the first to bring our attention to its importance. Like Keynes, Peirce was keenly aware of the distinction between the balance and the weight of the evidence. While Peirce develops the ideas on behalf of the Bayesian of his time in order to demonstrate its shortcoming, Peirce brings about formal develop of Bayesian ideas that come to be influential. His development of the notion of the balance of the

¹Charles S. Peirce, "The Probability of Induction," in Writings of Charles S. Peirce: A Chronological Edition, Volume 3: 1872–1878 (Indiana University Press, 1986).

²Technically they were followers of Laplace, who developed many Bayesian ideas.

evidence is especially important. We already discussed Keynes' view on the notion, which is identified as the magnitude of a probability. Peirce's idea is a bit more involved—this is partly due to his concern with the viability of a based epistemology based on partial belief, while Keynes simply assumes a Russellian epistemology that serves as the philosophical springboard for his interpretation of probability.

Peirce takes that the balance of the evidence as an *epistemological* thesis about the justification of belief that has its root in Human epistemology.³ In *An Enquiry Concerning Human Understanding*, Human makes his famous argument against the existence of miracle by appealing to the concept of probability. Human argues that due to the fact that the only evidence we have regarding miracle is witness testimony, whether or not miracle exists can only be determined probabilistically by determining the balance of the evidence:

All probability, then, supposes an opposition of experiments and observations, where the one side is found to overbalance the other, and to produce a degree of evidence, proportioned to the superiority⁴

The idea is that probability should allow us to combine evidence for both sides of the argument, and determine which is favored by all the evidence we have. In addition, Hume thought that the balance of the evidence is not just a comparative measure, but also a quantitative one:

In all cases, we must balance the opposite experiments, where they are opposite, and deduct the smaller number from the greater, in order to

³Charles S. Peirce, *The Essential Peirce, Volume 2: Selected Philosophical Writings (1893-1913)*, ed. Peirce Edition Project (Indiana University Press, 1998), 76.

⁴David Hume, An Enquiry Concerning Human Understanding and Other Writings (Cambridge University Press, 2007), 97.

know the exact force of the superior evidence.⁵

Peirce thought that the best case for Bayesianism is to combine Hume's epistemology with the probability theory developed by Laplace. Peirce proposes a measurement of the belief H on evidence E, we should take the odds of the probability of H conditional on E versus its probability conditional on $\neg E$. In other words,

$$\frac{P(H|E)}{P(\neg H|E)}$$

Which can be algebraically decomposed into

$$\frac{P(E|H)P(H)}{P(E|\neg H)P(\neg H)}$$

Using Principle of Indifference, which will be discussed below, the above becomes $\frac{H}{\neg H} = 1$, the balance of evidence is simply $\frac{P(E|H)}{P(E|\neg H)}$, which is an expression of what we now call the Bayes Factor. Peirce then proposes that Hume's idea can be captured by taking the log of the ratio. Consider the following expression:

$$\frac{P(E|H)P(H)}{P(E|\neg H)P(\neg H)} = log(P(E|H)) - log(P(E|\neg H))$$

A positive value means the evidence is in favor of H and negative against it. It also captures Hume's idea that the strengths of two independent pieces of evidence should be combinable. Suppose E_1 and E_2 are independent. So, $P(E_1 \wedge E_2|H) = P(E_1|H)P(E_2|H)$. If we take the logarithm of the product, the product becomes $log(P(E_1|H)) + log(P(E_2|H))$. This captures the idea that adding two pieces of

⁵Hume, 98.

independent evidence together should increase the intensity of our belief. Some modern Bayesians were so impressed with Peirce's account that they incorporated Peirce's ideas into their own frameworks. I. J. Good refers to Peirce as a precursor of his account.⁶ Branden Fitelson cites Peirce as the inspiration of his Bayesian account of independent evidence.⁷

However, Peirce argues that the balance of the evidence alone cannot be satisfactory capture our state of belief. Peirce illustrates the idea roughly as follows: imagine two urns A and B with unknown proportions of black and white balls. Suppose you sample (with replacement) 100 balls from the urn A and find 50 black balls and 50 white balls. Justifiably, you infer that the proportion of black balls in A - call it θ_A is about 0.5. You then decide to sample from B, but this time you only manage to draw 4 samples, 2 of which are black balls. Your best estimate for θ_B is 0.5. At this point, I offer you another chance to draw from one of the urns, and if you manage to draw a black ball from that urn, you get \$100. Which urn would you pick?

Clearly, $\theta_B = \theta_A$, but it is unclear if we should regard them with indifference, for a simple reason: the evidence for $\theta_A = 0.5$ is more substantial than the evidence for $\theta_B = 0.5$.

What Peirce has in mind, it should be clear, is precisely what Keynes later calls the weight of evidence. However, while Keynes was hesitant in affirming the necessity of evidential weight, Peirce thinks its role is indispensable:

to express the proper state of our belief, not one number but two are

⁶I. J. Good, "An Error by Peirce Concerning Weight of Evidence," *Journal of Statistical Computation and Simulation* 13, no. 2 (1981): 155–57.

⁷Branden Fitelson, "A Bayesian Account of Independent Evidence with Applications," *Proceedings* of the Philosophy of Science Association 2001, no. 3 (2001): S123.

requisite, the first depending on the inferred probability, the second on the amount of knowledge on which that probability is based.⁸

Unfortunately, however, this is the only instance of Peirce's explicit discussion of the weight of evidence. It would seem that the matter would not be broached again until Keynes' discussion in his *Treatise*, which comes to entirely overshadow Peirce's discussion of it. Even now, as Peirce scholars come to be interested in Peirce's discussion, it tends to be approached as a historical question of whether or how Perice has anticipated the later conceptions of evidential weight, such as Keynes' and Good's. However, what I will attempt is to construct an account of the weight of evidence within a largely Peircean framework. This will show that many puzzles regarding the weight of evidence that arise from Keynes' understanding can be resolved with Peirce's epistemology of inquiry.

3 Abduction

[in progress]

An important key to a Peircean interpretation of the weight of evidence is Peirce's tripartite classification of reasoning. As is well known, Peirce is responsible for coining the phrase "abduction", which is distinguished from deduction and induction. Peirce's understanding deduction and induction is very close to the modern understanding of the distinction. Deduction considers the necessary consequences between propositions,

⁸Peirce, "The Probability of Induction," 296.

⁹Jeff Kasser, "Two Conceptions of Weight of Evidence in PeirceIllustrations of the Logic of Science," *Erkenntnis* 81, no. 3 (2016): 629–48.

so a derivation in, say, predicate logic and theorem proving would be considered as deduction by Peirce. Induction, on the other hand, involves inference that is neither necessary nor justified by the relation between propositions, but by its reliability and conduciveness to knowledge. In scientific context, deduction is supposed to tease out the logical consequences of a hypothesis, which are then treated as predictions to be tested using inductive methods.

Where does abduction come in then? abduction is "the first starting of a hypothesis and the entertaining of it." This must be understood in the context of the context of Peirce's epistemology: inquiry—the practice of generating knowledge—is an interplay between doubt and belief. Doubt and belief are distinguished by their psychological and practical differences: psychologically, doubt is an uneasy and unstable state, while belief is characterized by imperturbability—a sense of stability of the mind. Practically, a belief expresses itself as as a stable habit—an established second nature, while the instability of doubt prompts us to seek satisfaction through reaching a state of belief. The degree of a belief, in turn, expresses itself in terms of the decisiveness of action. 12

Thus, Aduction is a reaction to the irradiation of doubt, caused by the perception of some surprising facts.¹³ It is an inferential process in which a hypothesis is proposed as a possible account for the the surprising fact. Thus, the role of abduction is

¹⁰Charles S. Peirce, Collected Papers of Charles Sanders Peirce (Cambridge: Harvard University Press, 1931), 6.525.

¹¹Charles S. Peirce, "The Fixation of Belief," in Writings of Charles S. Peirce: A Chronological Edition, Volume 3: 1872–1878 (Indiana University Press, 1966), 247.

¹²This is not to say Peirce believes in subjective probability. While Peirce has indicated that beliefs can be measured quantitatively, as far as the interpretation of probability goes, he is an objective propensity theorist.

¹³Peirce, Collected Papers of Charles Sanders Peirce, 1931, 5.512.

paramount to the growth of knowledge, as it is "the only logical operation which introduces new ideas." For an analogous position, consider Thomas Kuhn's idea that a shift in scientific paradigm has to be initiated by the perception of some anomalous event: Roentgen's the discovery of X-rays, instance, was caused by his perception of the surprising fact that his barium platinocyanide screen was unexpectedly glowing. ¹⁵

Peirce often uses the following syllogistic schema to illustrate the inferential process of abduction:

- 1. The surprising fact, C, is observed;
- 2. But if A were true, C would be a matter of course.
- 3. Hence, there is reason to suspect that A is true. 16

In this schema, C is the observed fact that calls for a hypothesis that would, in Peirce's words, rationalize it.¹⁷ Peirce sometimes refers to A as an abductive suggestion that follows the perception of the surprising fact.¹⁸ They are possible accounts for the phenomenon observed, and are not presented as true proposition, but hypotheses that we may accept provisionally for the sake of further testing. Thus, in abduction we infer that a particular hypothesis is worthy of investigation.

How, then, does abductive reasoning guide us in coming up with productive hypothesis that accounts for the facts observed? If abduction is to be a species of reasoning, on par with deduction and induction, there has to be some principled way to distinguish good abductions from the bad ones. Yet Peirce's characterization of

¹⁴Peirce, The Essential Peirce, Volume 2, 1998, 216.

 $^{^{15} \}text{Thomas S.}$ Kuhn and Ian Hacking, The Structure of Scientific Revolutions: 50th Anniversary Edition (University of Chicago Press, 2012), 57–58.

¹⁶Peirce, Collected Papers of Charles Sanders Peirce, 1931, 5.189.

¹⁷Peirce, The Essential Peirce, Volume 2, 1998, 107.

¹⁸Peirce, 227.

abduction sometimes the process almost seems either magical or random:

The abductive suggest comes to us like a flash. It is an act of insight, although of extremely infallible insight.¹⁹

... abduction is, after all, nothing but guessing.²⁰

However, just because abduction is ultimately guessing does not mean that it involves nothing but luck. An important idea is that the beginning of abduction usually involves a leap of faith, but this leap can be made strategically. Peirce explains the strategic nature of abduction with an analogy of the game of "twenty questions", in which one party has to think of an object, while another party has to find out what the object by asking 20 or less questions.²¹ Peirce's idea is that posing a question in the game is akin to proposing a hypothesis in an inquiry—for both cases, we make a guess about the nature of the subject of our inquiry, and the difference is that in the game the answer we receive is certain and direct, while a hypothesis has to be tested or broken down into sub-hypothesis. Peirce has an important point in mind, however, with this example. That is, the choice of a question can drastically alter the course of an inquiry. Since the question of the game has to figure out what the answer is by asking at most 20 questions, she ask them strategically. Each question the inquire asks should be as informative as possible—they should, for instance, ask questions that narrow the space of possibilities as quickly as possible. Abduction, which is the logic of hypothesis selection, is guided by a similar goal.

Hintikka suggests that looking at the role played by strategic thinking in deductive

¹⁹Peirce, 227.

²⁰Peirce, Collected Papers of Charles Sanders Peirce, 1931, 7.219.

 $^{^{21}???}$

thinking is helpful in understanding what Peirce has in mind. Hintikka points out that even though there is an emphasis on mechanical rules of inference, deductive thinking involves more than that. Suppose a student is asked, in an exam for a logic class, to demonstrate that

$$\vdash [[A \to (B \land \neg C)] \land (\neg B \lor D)] \to (A \to D)$$

In other words, to prove that it is a logical truth. Before proving that, she must make a decision on which method to use. Since she is pressed for time, she must choose to prove this strategically. For instance, suppose she choose the strategy of trying to show that it is always true. one way would be to use the truth table to show that the proposition is true in all possible interpretations. But that would be a inefficient choice, since there will be $2^4 = 16$ rows and she cannot skip any row since she's trying to prove a logical truth. Alternatively, she could prove it using natural deduction, which calls more further strategic thinking: an often recommended strategy for proving logical truth is by $Reductio\ ad\ Absurdum$, that is, by showing a contradiction from the negation of the said logical truth. But in this case, this would hardly seem like a good decision, since she would end up with a long negated conditional. A better strategy would be to prove the result conditionally assuming the antecedent of the conditional. Depending on the rules available to her, such a method would take 10 or less lines. So, a successful deductive reasoner must know all the basic rules of inferences, but she must also master policies or strategies that are more suggestive in nature.

The important point here is that the strategic dimension of reasoning often finding a way to gain as much information as possible in light of some constraint. In the game of twenty questions, the constraint is that the questioner only gets 20 questions. For the student taking a logic exam, she is bound by the time she has. In both cases, the reasoner has to make a decision on confronting the problem she faces in an efficient manner. In particular, before even tackling the problem they are interested, both reasoners must select one of many methods in which the problem could be addressed. Often, the efficacy of the chosen method cannot be determined until it has been actually tried, so there will be always be an element of guessing involved.

The strategic dimension of abduction, as Peirce sees it, pertains to how our resources is allocated in service to our epistemological goal, and this is addressed by a branch of abductive reasoning called the economy of research. The creative and suggestive nature of abductive inference makes it permissive in nature, and, for it to be reliable, it must be guided by economical considerations that discriminate between productive and unproductive proposals. According to the schema Peirce provides, we can infer that there is a reason to suspect a hypothesis to be true, as long as it accounts for the facts somehow. To be sure, Peirce makes it clear that the permissive nature of abduction must be tempered by the pragmatic maxim—a hypothesis must be distinguishable by their experimental and practical consequences.²² Still, for many facts, there could be countless possible testable accounts, and examining every single possible abductive suggestion is not only imprudent, but it impedes greatly our ability to find the right hypothesis.²³ Thus, how to deal with this problem, in light of our limited resources, is the heart of the economy of research and the rationality of abduction:

²²Peirce, The Essential Peirce, Volume 2, 1998, 234.

²³Peirce, Collected Papers of Charles Sanders Peirce, 1931, 2.776.

Proposals for hypotheses inundate us in an overwhelming flood, while the process of verification to which each one must be subjected before it can count as at all an item, even of likely knowledge, is so very costly in time, energy, and money—and consequently in ideas which might have been had for that time, energy, and money, that Economy would override every other consideration even if there were any other serious considerations. In fact there are no others.²⁴

Thus, even though

The key here is to see abduction both as a *creative* and *rational* procedure. On one hand,

Explanatory virtue is obviously one guide—the proposed hypothesis must explain the fact in some way

Hintikka's suggestion

strategic

My claim is that the weight of evidence is relevant not in the inductive stage of inquiry, but in the abductive stage. To begin,

More important, abduction

First, the weight of evidence pertains to the *strategic* stage of abduction

Example: Recall one of Richard Jeffrey's responses to Popper's paradox of ideal evidence is that the prior we choose affects how we learn.

²⁴Peirce, 5.602.

illustration: generate random sample

Second, we saw that Skyrms's idea of resiliency is the measure of a probability's relative stability against a certain surprising

4 economy of resarch

For Peirce, the weight of evidence is tightly connected to the notion of the precision of a probability, as he suggests it should be based on what he calls probable error, which is a precursor to the modern Frequentist notion of confidence interval.²⁵ This is evidenced by Peirce's example above—it anticipated what is now known as the Ellsburg Paradox, an empirical phenomenon in which human subjects are shown to be sensitive to difference between precise(a point value) and imprecise(an interval) probabilities.²⁶ Peirce was impressed by Boole's way of deriving indefinite probabilities by using arbitrary constants.²⁷ Boole's view is seen as the precursor of modern theory of imprecise probability.²⁸ It is then no surprise that Peirce is cognizant of the nuance issues surrounding the precision of probability.

Peirce, furthermore, has a dynamic view on the weight of evidence. Unlike Keynes, Peirce does not seem to think the weight of evidence *always* increase when new relevant evidence is introduced; because, he thinks that the weight of evidence

²⁵Ian Hacking, *The Taming of Chance* (Cambridge University Press, 1990).

²⁶Daniel Ellsberg, "Risk, Ambiguity, and the Savage Axioms," *The Quarterly Journal of Economics* 75, no. 4 (1961): 643–69.

²⁷Charles S. Peirce, "Harvard Lecture Vi, 1865," in Writings of Charles S. Peirce: A Chronological Edition, Volume 1: 1857–1866 (Indiana University Press, 1982), 239.

²⁸Peter Walley, Statistical Reasoning with Imprecise Probabilities (Chapman & Hall, 1991), 43.

is tied to the probability's "liability to be changed by further experience." So—to use the terminology from chapter 1—when our belief is resilient, the weight of the evidence will not change even if new evidence is introduced.

5 Moore's Anti-Utilitarian Argument and The Problem of Total Evidence

Peirce's insight is closely related to a puzzle between evidential weight and expected utility in A Treatise in Probability. In chapter 1, we saw that Keynes' epistemology was squarely in the Moore-Russell tradition. Keynes' notion of probability as intuitable logical relation is influenced by Moore moral intuitionism, a position that takes 'good' to be an indefinable but intuitable property of an object.³⁰ Moore is, in particular, well known for his criticism of utilitarianism for committing what he calls the "naturalistic fallacy" The idea is that since 'good' is a basic unanalyzable property, it would be a mistake to, as utilitarians do, explicate it as another natural property such a pleasure or happiness.

This, of course, flies in the face of any mathematical analysis of the right action, which requires a quantitative analysis of the amount of good generated by an action. More important, such an approach often requires the use of probability, since an action that produces some good with a low probability ought not be the same as one that generates the same degree of good but with a much higher probability. This is

²⁹Peirce, "The Probability of Induction," 295.

³⁰G. E. Moore, *Principia Ethica* (Prometheus Books, 1988), 9–10.

 $^{^{31}}$ Moore, 11–12.

expressed clearly by Pascal and his fellow Jansenists in the *Port Royal Logic* in the 17th century:

...in order to decide what we ought to do to obtain some good or avoid some harm, it is necessary to consider not only the good or harm in itself, but also the probability that it will or will not occur, and to view geometrically the proportion all these things have when taken together.³²

Moore's naturalistic fallacy can be seen as an attack on the use of the mathematical analysis in ethics, but he also has an argument against the use of probability in ethical reasoning. This deeply concerns Keynes. Moore's argument is offered in the chapter "Ethics in Relation to Conduct." His argument is that when we analyze the the good or utility generated by an action, we often only take into account the probability of the immediate utility that would follow from the action, but to take utilitarianism seriously, Moore claims, we must consider the effects the action in the long the run.³³ We have to know, for instance, whatever good our action produces in the immediate future will not be negated by the negative effects it has in the long run. However, Moore argues that, because so much about the future is currently unknown to us, we simply lack the ground to make good inferences about what effects each single action will probably case in the long run. For a somewhat crude example, consider a trolley-type problem in which I have to choose between saving a doctor or a criminal. I may reason that saving the doctor will more probably produce more utility in the future, since she will likely save more lives, but this only reflects the short-term outcomes. Why shouldn't I consider, for instance, the probability of the doctor engaging in malpractice and

³³Moore, Principia Ethica, 158.

³²Antoine Arnauld, Logic, or, the Art of Thinking: Containing, Besides Common Rules, Several New Observations Appropriate for Forming Judgment (Cambridge University Press, 1996), 273.

causing a significant amount of suffering, or the probability that the criminal might turn her life around and become a productive member of the society?

5.1 Challenge 1:

Keynes resists this line of argument. To begin, Keynes takes Moore's argument to have demonstrated at most that we cannot know the long term consequence of our action with certainty, but this sets the bar too high—the whole point of employing probabilistic reasoning in our thinking about the right conduct is to maintain a degree of rationality despite of our ignorance about the future. A probability that is based on very little evidence is still "a genuine probability."³⁴

Moore's conclusion, as Keynes points out, does make sense from a Frequentist point of view. Most, if not all, of our actions cannot be understood as a part of a long term frequency, so naturally the use of probability to determine the right conduct has a very little meaning from this perspective—if probability is strictly defined as something that expresses itself only in long term behavior, we cannot speak of the probability of an event with any credibility unless we can study the event in a controlled and repeatable environment.

Keynes' logical interpretation, however, does not require the meaning of a probability to be grounded in an empirically verified frequency. A probability is meaningful as long as we rightly perceive the logical relation between the premises and the conclusion. If we have no reason to think that our action is more likely to cause one long term consequence than another, then, on Keynes' interpretation of probability, we

³⁴John Maynard Keynes, A Treatise on Probability (Macmillan; Co., Limited., 1921), 354.

should rightly regard these events with indifference.³⁵

5.2 Challenge 2 WOE

Keynes, however, is unwilling to wholeheartedly endorse the doctrine of the maximization of utility—he is especially concerned with the fact that expected utility does not account for the weight of evidence, since utility is only discounted by the magnitude of its probability. So as far as expected utility theorists are concerned, if the two probabilities have the same value, it makes no practical difference if the probability is based on the Principle of Indifference, or a big dataset that is uniformly distributed.

However, Keynes points out that proponents of expected utility theory often does implicitly assume the importance of the weight of evidence. Bernoulli, for instance, suggests that rationality demands the utilization of all evidence available to us. This implies, Keynes thinks, that it's always rational to get more evidence, but then it raises another critical question about whether or not one could ever be rational in refusing new evidence.³⁶ If the answer for the former question is positive, and the latter question negative, then we have to conclude that rationality dictates us that we should never stop looking for more evidence.

Keynes does not make the jump from "using all the evidence" to "get all the evidence" clear. Nevertheless, this problem is revisited many years later in an exchange between Ayer and Carnap. So drawing a parallel between them could be helpful.

In his Logical Foundation of Probability, Carnap restates Bernoulli's maxim as

³⁵Keynes, 354.

 $^{^{36}}$ Keynes, 84–85.

"the requirement of total evidence".

Requirement of total evidence: in the application of inductive logic to a given knowledge situation, the total evidence available must be taken as basis for determining the degree of confirmation.³⁷

Aver, in response to Carnap, raises the Keynesian question: should "total evidence" include relevant evidence that I do not yet have in possession?³⁸ The answer must be "yes", Ayer argues. If finding the truth value of some proposition P could potentially sway the balance of my evidence, then I should definitely acquire it. Thus the principle of total evidence seems to suggest that I am also rationally compelled to consider some evidence I do not yet have.

But Ayer points out that this cannot be the whole picture: taken as a rule of rationality, this means we should never stop acquiring unless we are certain that we have acquired all available evidence. This, however, assumes that we know what evidence is available, but it could often be unrealistic to expect to know how much evidence we *do not* currently have.

Ayer argues that this is only a symptom of a deeper problem about logical nature of probability and its bearing on the rationality of our evidential practice. The logical interpretation of probability, held by both Carnap and Keynes, takes probability as a logical relation between propositions, so within this picture, inductive rationality is a matter of having the right degrees of belief between premises and conclusion—this is analogous to the idea deductive rationality is perceive correctly whether the conclusion

 $^{^{37} \}mathrm{Rudolf~Carnap}, \ Logical~Foundations~of~Probability~(Chicago]$ University of Chicago Press, 1950), 211.

³⁸A. J. Ayer, *Probability and Evidence* (Macmillan, 1972), 56.

follows necessarily from the premises.

Logic, however, does not care about how much evidence we have; it only cares about the relation between our propositions. This problem affects the subjective interpretation of probability as well. As Leonard Savage points out, probability could reveal to us the incoherence within the web of our belief, but it cannot tell us how to resolve it.

According to the personalistic view, the role of the mathematical theory of probability is to enable the person using it to detect inconsistencies in his own real or envisaged behavior. It is also understood that, having detected an inconsistency, he will remove it. An inconsistency is typically removable in many different ways, among which the theory gives no guidance for choosing.³⁹

We encountered a version of this problem in chapter 1, in which we considered Keynes' definition of the weight of evidence in terms of the conditional relevance. Once again, the weight of evidence seems directly relevant to inductive reasoning, yet it cannot be easily situated in the probabilistic framework. The notion of resiliency, which we discussed in detail in chapter 1, does not seem to do any better. While it captures the expression of the weight of evidence, it does very little in illuminating on how it dedicates the rationality of our decisions. Both Joyce and Skyrms are silent on this.

³⁹Leonard J. Savage, *The Foundations of Statistics* (Dover, 1954), 57.

6 The Weight and Value of Evidence

Keynes, again, is rather ambivalent about resolving the tension between the weight of evidence and how inductive reasoning is generally conceived. In pass, however, he makes a tantalizing suggestion that the *utility* of evidence could be the component needed to resolve the situation. He begins with the observation that getting evidence for a belief low in evidential weight will "probably produce the greatest amount of good", but the situation is opposite when the evidence for the belief is weighty—"there clearly comes a point when it is no longer worth while to spend trouble" Thus, for a hypothesis of interest H, the same evidence E generates different amount of utility relative to the amount of information we already have for H. If an agent has almost no information about H, gathering more information would generate the most utility, but for the same evidence, the demand might to low, because the agent might already have enough information about H, so getting more evidence would yield very little to no utility.

This opens the door to another conception of the weight of evidence, one that is not purely based in the *amount* of evidence, but also on how much how much we want. In the *Treatise*, Keynes does switch implicitly these two way of thinking of evidential weight—sometimes refers the weight of evidence a balance "between the absolute amounts of relevant knowledge and of relevant ignorance respectively" In a later chapter of the *Treatise*, he also calls weight "the degree of completeness of the information" These remarks suggest that weight is about how much evidence we

⁴⁰Keynes, A Treatise on Probability, 84–85.

⁴¹Keynes, 78.

⁴²Keynes, 357.

desire as much as how much we do know. This also allows the possibility that new relevant evidence does not always increase the weight of the evidence, because if I already have enough evidence, getting more might not always be informative.

The puzzle about the utility of evidence, and its bearing on the rationality of the gathering of evidence, has been addressed by Ramsey on an unpublished note. Interestingly, however, he has proven essentially the opposite conclusion reached by Ayer and Keynes: Ramsey shows that we should always look more more evidence, because we can never be worse off from doing it. How can this be?

Ramsey's argument is roughly that, if we assume an agent to be a perfect Bayesian and that new information does not cost anything, then she will never be no worse off getting new evidence.⁴³ In fact, she is guaranteed to be *better* off as long as the new evidence will tell her something new. A perfect Bayesian agent is someone who studiously updates her opinions based on Bayes' rule and then act by choosing the action that maximize her expected utility. Note that this assumes two things: first, for any decision problem she faces, there is always going at least one course of action that maximizes her expected utility, and second, as Skyrms points out, this also implies that the agent knows that she will always stays being perfectly Bayesian in the future.

I will make use of an intuitive example rather than reproducing the proof here.⁴⁴ Suppose we have three hypotheses about the content of an urn in front of us:

⁴³F. P. Ramsey, "Weight or the Value of Knowledge," *British Journal for the Philosophy of Science* 41, no. 1 (1990): 1–4, also see @goodtotalevidence and @savage, sec 6.2.

⁴⁴This example is adapted from Isaac Levi, "The Weight of Argument," in *Fundamental Uncertainty:* Rationality and Plausible Reasoning, ed. Silva Marzetti Dall'Aste Brandolini and Roberto Sczzieri (Palgrave MacMilan, 2011), 39–58

- 1. H_b : 90 black balls and 10 white balls
- 2. H_w : 10 white balls and 90 black balls
- 3. H_n : 50 white balls and 50 black balls.

Suppose we start by assuming $P(H_b) = P(H_w) = P(H_n) = 1/3$ —we could have some knowledge that assures us that these are the only three possibilities. There is also a reward of \$1 for picking the correct hypothesis. Our expected payoff for choosing each hypothesis would be the same at 1/3. Nevertheless, we are allowed to sample with replacement as many times as we wish. Should we get more evidence? Yes, according to Ramsey, we should, and this can demonstrated in terms of an expected utility analysis.

To begin, at this point, the probability of getting a black ball is the same as getting a white ball. Let E_b be "a black ball is drawn" and E_w for white balls. So:

$$P(E_b) = P(H_b)P(E_b|H_b) + P(H_w)P(E_b|H_w) + P(H_n)P(E_b|H_n)$$
$$= 1/3(0.9) + 1/3(0.1) + 1/3(0.5) = 0.5$$

And $P(E_w) = 1 - P(E_b) = 0.5$. So, in the event of drawing a black ball from the urn, we would update our belief like so:

$$P(H_b|E_b) = \frac{P(H_b)P(E_b|H_b)}{P(E_b)} = \frac{1/3(0.9)}{0.5} = 0.6$$

Similarly, applying the calculation on the other hypotheses, we get:

$$P(H_w|E_b) = 0.067$$

$$P(H_n|E_b) = 0.333$$

Similar argument can be made by assuming E_w , that is, a white ball is chosen. In that case $P(H_w|E_w) = 0.6$. If we were an ideal Bayesian agent, we should pick H_b if E_b , and pick H_w if E_w . Since an ideal Bayesian would choose the option that maximizes our expected utility, in either case the expected value after drawing from the urn once is 0.6, which is an improvement, since before drawing our expected utility is 1/3 for all options. The net gain in expected utility would be 0.6 - 0.33 = 0.27, is referred to as the value of information in the decision theory literature.⁴⁵

It turns out that we would be even better off if we were to draw from the urn again. Suppose the first draw yields a black ball. So now we have one piece of evidence in hand. Let us refer to our state of belief after the first draw as H'_b, E'_b , .. and so on. For instance, $P(H'_b) = P(H_b|E_b)$ and $P(E'_b) = P(E'_b|E_b)$. One notable change is that $P(E'_b) = 0.7132$ and $P(E'_w) = 0.2868$. If we draw again and get a black ball, this means:

$$P(H_b'|E_b') = 0.757$$

$$P(H_w'|E_b') = 0.009$$

$$P(H_n'|E_h') = 0.233$$

If a white ball were to be drawn:

 $^{^{45} \}rm Howard$ Raiffa and Robert Schlaifer, Applied Statistical Decision Theory (Harvard, 1964), 89–90. For a more digestible presentation see @winkler sec.6.3.

$$P(H_b'|E_w') = 0.21$$

$$P(H_w'|E_w') = 0.21$$

$$P(H_n'|E_w') = 0.58$$

Thus, if the second sample is a black ball, we would choose b since it has the maximum expected utility at 0.757, and if we get a white ball, we choose n with the expected value at 0.58. So, the expected utility, if we were to draw from the urn again, is: 0.7132(0.757) + 0.2867(0.58) = 0.706, which is an improvement over just drawing once. The net gain is 0.706 - 0.6 = 0.106. Ramsey's proof shows that we can keep on getting more evidence and we will never be worse off. In fact, we will be better off as long as there is evidence out there we do not yet have.

What should we make of Ramsey's proof? There are two issues involved here. The first is Keynes' observation that evidence can have a diminishing return, so relevant evidence does not increase weight, and the other is how the weight of evidence bears on the rationality of our action, especially when it comes to the gathering evidence. Ramsey's note provides a good answer for the former, but not the latter.

To begin, Ramsey's contribution here is a way for us to think about the relationship between the weight of the evidence in possession and the value of the potential new evidence.⁴⁶ Ramsey clearly thought is the value of evidence E for hypothesis H as something along the line of the difference between the prior expected utility EU(H)

⁴⁶As noted, this is essentially the idea of the value of information in decision theory, but, as a historical note, Ramsey, inspired by Keynes' puzzle about the weight of evidence, has anticipated this development by many years.

and the posterior EU(H|E).

For instance, we saw that in the example above, the posterior expected utility of the first draw was 0.27 higher than our prior expected utility, and we saw a net gain of 0.106 in expected utility if we were to draw again after drawing a black ball, so we see that the first piece of evidence has a higher value than the second one. What Ramsey's proof demonstrates is that new evidence has a diminishing return—I get a "bigger bang for the buck" for my evidence gathering endeavor when I have less evidence. In light of this, Keynes' example of the balance of the evidence unchanged by the introduction of new evidence is then somewhat incomplete. This explains one of Keynes' puzzle about worthiness of our endeavor to get more evidence in light of the evidence we have in possession.

However, the broader normative question is still unanswered: how should the weight of evidence guide the rationality of our action? To be sure, I do not question that given some assumptions, Ramsey's result will necessarily follow: the same result is proven by both Good and Leonard Savage, so there is no doubt that the result will holds if the assumptions are granted, but that's a big *if*—we have to question if how often these assumptions actually hold.

The scenario we imagined quickly breaks down once we starts to introduce some sort of cost. It was assumed in the example that it costs us neither money nor time to draw from the urn, but suppose it costs us 25 cents for each sample. This means that we would be gaining only 0.27 - 0.25 = 0.02 in expected payoff for the first draw, and the second draw would definitely not be worth the additional 25 cents. Or suppose that one dollar is not worthy any endeavor that lasts longer than 15 seconds, and it

takes 30 seconds to draw from the urn. As soon as minimally realistic assumptions are introduced, Ramsey's result no longer holds.

Cost might also enter into consideration in different forms, e.g., computational cost or memory. In the same context, Savage ponders over an interesting case that introduces yet another dimension of the problem: consider a very ill person, who is given the option to find out with no cost if the disease she has is mortal. Savage points out that an argument can be made that in this case refusing information could be rational. The thought is that the patient may decide that, based on an assessment of her own personality, she would live the rest of her remaining life in agony if she were to find out that her disease is very serious, whereas she could live relatively happily without knowing. Savage's response is that in this case the information is not really free; it has a psychological cost.⁴⁷

Savage's response was intended to paint a counterexample as an explainable exception to the idea that it's always rational to get free evidence, but this sort of implicit cost is the norm, not an exception: at the minimal, any endeavor to seek more evidence will cost us at least time, and the loss of time is the loss of opportunity. Cost might also enter into consideration in different forms, e.g., computational cost or memory. In light of this, Ramsey's demonstration—the value of evidence depends on how much evidence we have and how much evidence there is to get—does not ease Keynes' concern at all, but it makes the question of evidence gathering more pressing: I might not be getting the most out of my evidential endeavor if it turns out that I could be much better off by examining a different hypothesis, for the very set up for Ramsey's proof requires us to have already chosen *one* hypothesis of interest. But in

⁴⁷Savage, The Foundations of Statistics, 107.

reality, we often have competing interests. The question is not whether I will get some utility out of gathering evidence for hypotheses in which I have interest, but how I should go about doing it, given the limitation of resource.

7 Epistemic Agency

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