Statistical Explanation

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EVER SINCE HIS CLASSIC PAPER with Paul Oppenheim, “Studies in the Logic of Explanation,” first published in 1948, Carl G. Hempel has maintained that an “explanatory account [of a particular event] may be regarded as an argument to the effect that the event to be explained . . . was to be expected by reason of certain explanatory facts” (my italics). It seems fair to say that this basic principle has guided Hempel’s work on inductive as well as deductive explanation ever since. In spite of its enormous intuitive appeal, I believe that this precept is incorrect and that it has led to an unsound account of scientific explanation. In this paper I shall attempt to develop a different account of explanation and argue for its superiority over the Hempelian one. In the case of inductive explanation, the difference between the two treatments hinges fundamentally upon the question of whether the relation between the explanans and the explanandum is to be understood as a relation of high probability or as one of statistical relevance. Hempel obviously chooses the former alternative; I shall elaborate an account based upon the latter one. These two alternatives correspond closely to the “concepts of firmness” and the “concepts of increase of firmness,” respectively, distinguished by Rudolf Carnap in the context of confirmation theory. Carnap has argued, convincingly in my opinion, that confusion of these two types of concepts has led to serious trouble in inductive logic; I shall maintain that the same thing has happened in the theory of explanation. Attention will be focused chiefly upon inductive explanation, but I shall try to show that a similar difficulty infects deductive explanation and that, in fact, deductive explanation can advantageously be considered as a special limiting case of inductive explanation. It is my hope that, in the end, the present relevance account of scientific explanation will be justified, partly by means of abstract “logical” considerations and partly in terms of its ability to deal with problems that have proved quite intractable within the Hempelian schema.

1. The Hempelian Account

Any serious contemporary treatment of scientific explanation must, it seems to me, take Hempel’s highly developed view as a point of departure. In the famous 1948 paper, Hempel and Oppenheim offered a systematic account of deductive explanation, but they explicitly denied that all scientific explanations fit that pattern; in particular, they called attention to the fact that some explanations are of the inductive variety. In spite of fairly general recognition of the need for inductive explanations, even on the part of proponents of Hempel’s deductive model, surprisingly little attention has been given to the problem of providing a systematic treatment of explanations of this type. Before 1965, when he published “Aspects of Scientific Explanation,” Hempel’s “Deductive-Nomological vs. Statistical Explanation” was the only well-known extensive discussion. One could easily form the impression that most theorists regarded deductive and inductive explanation as quite similar in principle, so that an adequate account of inductive explanation would emerge almost routinely by replacing the universal laws of deductive explanation with statistical generalizations, and by replacing the deductive relationship between explanans and explanandum with some sort of inductive relation. Such an attitude was, of course, dangerous in the extreme, for even our present limited knowledge of inductive logic points to deep and fundamental differences between deductive and inductive
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This fact should have made us quite wary of drawing casual analogies between deductive and inductive patterns of explanation. Yet even Hempel's detailed examination of statistical explanation may have contributed to the false feeling of security, for one of the most significant results of that study was that both deductive and inductive explanations must fulfill a requirement of total evidence. In the case of deductive explanations the requirement is automatically satisfied; in the case of inductive explanations that requirement is nontrivial.

Accordingly, the situation in May 1965, at the time of the Pittsburgh Workshop Conference, permitted a rather simple and straightforward characterization which would cover both deductive and inductive explanations of particular events. Either type of explanation, according to Hempel, is an argument; as such, it is a linguistic entity consisting of premises and conclusion. The premises constitute the explanans, and the conclusion is the explanandum. The term "explanandum event" may be used to refer to the fact to be explained; the explanandum is the statement asserting that this fact obtains. The term "explanatory facts" may be used to refer to the facts adduced to explain the explanandum event; the explanans is the set of statements asserting that these explanatory facts obtain. In order to explain a particular explanandum event, the explanatory facts must include both particular facts and general uniformities. As Hempel has often said, general uniformities as well as particular facts can be explained, but for now I shall confine attention to the explanation of particular events.

The parallel between the two types of explanation can easily be seen by comparing examples; here are two especially simple ones Hempel has offered:

(1) Deductive
This crystal of rock salt, when put into a Bunsen flame, turns the flame yellow, for it is a sodium salt, and all sodium salts impart a yellow color to a Bunsen flame.

(2) Inductive
John Jones was almost certain to recover quickly from his streptococcus infection, for he was given penicillin, and almost all cases of streptococcus infection clear up quickly upon administration of penicillin.

These examples exhibit the following basic forms:

(3) Deductive
All F are G.
z is F.
z is G.

(4) Inductive
Almost all F are G.
z is F.
2z is G.

There are two obvious differences between the deductive and inductive examples. First, the major premise in the deductive case is a universal generalization, whereas the major premise in the inductive case is a statistical generalization. The latter generalization asserts that a high, though unspecified, proportion of F are G. Other statistical generalizations may specify the exact numerical value. Second, the deductive schema represents a valid deductive argument, whereas the inductive schema represents a correct inductive argument. The double line in (4) indicates that the conclusion "follows inductively," that is, with high inductive probability. Hempel has shown forcefully that (4) is not to be construed as a deduction with the conclusion that "z is almost certain to be G." By the time Hempel had provided his detailed comparison of the two types of explanation, certain well-known conditions of adequacy had been spelled out; they would presumably apply both to deductive and to inductive explanations:

(i) The explanatory argument must have correct (deductive or inductive) logical form. In a correct deductive argument the premises entail the conclusion; in a correct inductive argument the premises render the conclusion highly probable.

(ii) The premises of the argument must be true.

(iii) Among the premises there must occur essentially at least one lawlike (universal or statistical) generalization.

(iv) The requirement of total evidence (which is automatically satisfied by deductive explanations that satisfy the condition of validity) must be fulfilled.

Explanations that conform to the foregoing conditions certainly satisfy Hempel's general principle. If the explanation is deductive, the explanandum event was to be expected because the explanandum is deducible from the explanans; the explanans necessitates the explanandum. If the
explanation is inductive, it "explains a given phenomenon by showing that, in view of certain particular facts and certain statistical laws, its occurrence was to be expected with high logical, or inductive, probability." In this case the explanandum event was to be expected because the explanans confers high probability upon the explanandum; the explanatory facts make the explanandum event highly probable.

2. Some Counterexamples

It is not at all difficult to find cases that satisfy all of the foregoing requirements, but that certainly cannot be regarded as genuine explanations. In a previously mentioned paper I offered the following inductive examples:

(5) John Jones was almost certain to recover from his cold within a week, because he took vitamin C, and almost all colds clear up within a week after administration of vitamin C.

(6) John Jones experienced significant remission of his neurotic symptoms, for he underwent extensive psychoanalytic treatment, and a substantial percentage of those who undergo psychoanalytic treatment experience significant remission of neurotic symptoms.

Both of these examples correspond exactly with Hempel's inductive example (2) above, and both conform to his schema (4). The difficulty with (5) is that colds tend to clear up within a week regardless of the medication administered, and, I understand, controlled tests indicate that the percentage of recoveries is unaffected by the use of vitamin C. The problem with (6) is the substantial spontaneous remission rate for neurotic symptoms of individuals who undergo no psychotherapy of any kind. Before we accept (6) as having any explanatory value whatever, we must know whether the remission rate for psychoanalytic patients is any greater than the spontaneous remission rate. I do not have the answer to this factual question.

I once thought that cases of the foregoing sort were peculiar to inductive explanation, but Henry Kyburg has shown me to be mistaken by providing the following example:

(7) This sample of table salt dissolves in water, for it has had a dissolving spell cast upon it, and all samples of table salt that have had dissolving spells cast upon them dissolve in water.

It is easy to construct additional instances:

(8) John Jones avoided becoming pregnant during the past year, for he has taken his wife's birth control pills regularly, and every man who regularly takes birth control pills avoids pregnancy.

Both of these examples correspond exactly with Hempel's deductive example (1), and both conform to his schema (3) above. The difficulty with (7) and (8) is just like that of the inductive examples (5) and (6). Salt dissolves, spell or no spell, so we do not need to explain the dissolving of this sample in terms of a hex. Men do not become pregnant, pills or no pills, so the consumption of oral contraceptives is not required to explain the phenomenon in John Jones's case (though it may have considerable explanatory force with regard to his wife's pregnancy).

Each of the examples (5) through (8) constitutes an argument to show that the explanandum event was to be expected. Each one has correct (deductive or inductive) logical form—at least it has a form provided by one of Hempel's schemata. Each one has true premises (or so we may assume for the present discussion). Each one has a (universal or statistical) generalization among its premises, and there is no more reason to doubt the lawlikeness of any of these generalizations than there is to doubt the lawlikeness of the generalizations in Hempel's examples. In each case the general premise is essential to the argument, for it would cease to have correct logical form if the general premise were simply deleted. We may assume that the requirement of total evidence is fulfilled in all cases. In the deductive examples it is automatically satisfied, of course, and in the inductive examples we may safely suppose that there is no further available evidence which would alter the probability of John Jones's recovery from either his cold or his neurotic symptoms. There is nothing about any of these examples, so far as I can see, which would disqualify them in terms of the foregoing criteria without also disqualifying Hempel's examples as well.

Folklore, ancient and modern, supplies further instances that would qualify as explanatory under these conditions:

(9) The moon reappeared after a lunar eclipse, for the people made a great deal of noise, banging on pots and pans and setting off fireworks, and the moon always reappears after an eclipse when much noise occurs. (Ancient Chinese folklore)

(10) An acid-freak was standing at the corner of Broadway and Forty-second Street, uttering agonizing moans, when a passerby asked him what was the matter. "Why, man, I'm keeping the wild tigers away," he answered. "But there are no wild tigers
around here," replied the inquirer. "Yeah, see what a good job
I'm doing!" (Modern American folklore)

A question might be raised about the significance of examples of the
foregoing kind on the ground that conditions of adequacy are normally
understood to be necessary conditions, not sufficient conditions. Thus,
the fact that there are cases which satisfy all of the conditions of
adequacy but which are not explanations is no objection whatever to the
necessity of such conditions; this fact could be construed as an objection
only if the conditions of adequacy are held to be jointly sufficient. In
answer to this point, it must be noted that, although the Hempel-Oppen-
heim paper of 1948 did begin by setting out conditions of adequacy, it
also offered a definition of deductive-nomological explanation, that is,
explicit necessary and sufficient conditions. The deductive examples
offered above satisfy that definition. And if that definition is marred by
purely technical difficulties, it seems clear that the examples conform to
the spirit of the original definition.

One might, however, simply acknowledge the inadequacy of the early
definition and maintain only that the conditions of adequacy should stand
as necessary conditions for satisfactory explanations. In that case
the existence of a large class of examples of the foregoing sort would seem
to point only to the need for additional conditions of adequacy to
rule them out. Even if one never hopes to have a set of necessary
conditions that are jointly sufficient, a large and important set of exam-

ples which manifest a distinctive characteristic disqualifying them as
genuine explanations demands an additional condition of adequacy.
Thus, it might be said, the way to deal with the counterexamples is by
adding a further condition of adequacy, much as Hempel has done in
"Aspects of Scientific Explanation," where he enunciates the requirement
of maximal specificity. I shall argue, however, that this requirement
does not exclude the examples I have offered, although a minor emenda-
tion will enable it to do so. Much more significantly, I shall maintain that
the requirement of maximal specificity is the wrong type of requirement
to take care of the difficulties that arise and that a much more funda-
mental revision of Hempel's conception is demanded. However, before be-

ginning the detailed discussion of the requirement of maximal specificity,
I shall make some general remarks about the counterexamples and their
import—especially the inductive ones—in order to set the stage for a
more enlightening discussion of Hempel's recent addition to the set of
conditions of adequacy.

3. Preliminary Analysis

The obvious trouble with our horrible examples is that the "explanatory"
argument is not needed to make us see that the explanandum event
was to be expected. There are other, more satisfactory, grounds for this
expectation. The "explanatory facts" adduced are irrelevant to the ex-
planandum event despite the fact that the explanandum follows (deduc-
tively or inductively) from the explanans. Table salt dissolves in water
regardless of hexing, almost all colds clear up within a week regardless
of treatment, males do not get pregnant regardless of pills, the moon
reappears regardless of the amount of Chinese din, and there are no wild
tigers in Times Square regardless of our friend's moans. Each of these
explanandum events has a high prior probability independent of the
explanatory facts, and the probability of the explanandum event relative
to the explanatory facts is the same as this prior probability. In this sense
the explanatory facts are irrelevant to the explanandum event. The
explanatory facts do nothing to enhance the probability of the explan-
andum event or to make us more certain of its occurrence than we would
otherwise have been. This is not because we know that the fact to be
explained has occurred; it is because we had other grounds for expecting
it to occur, even if we had not already witnessed it.

Our examples thus show that it is not correct, even in a preliminary
and inexact way, to characterize explanatory accounts as arguments
showing that the explanandum event was to be expected. It is more
accurate to say that an explanatory argument shows that the probability
of the explanandum event relative to the explanatory facts is substan-
tially greater than its prior probability. An explanatory account, on this
view, increases the degree to which the explanandum event was to be
expected. As will emerge later in this paper, I do not regard such a
statement as fully accurate; in fact, the increase in probability is merely a
pleasant by-product which often accompanies a much more fundamental
characteristic. Nevertheless, it makes a useful starting point for further
analysis.

We cannot, of course, hope to understand even the foregoing rough
characterization without becoming much clearer on the central concepts
it employs. In particular, we must consider explicitly what is meant by
"probability" and "prior probability." There are several standard views
on the nature of probability, and I propose to consider briefly how the
foregoing characterization of explanation, especially inductive explana-
tion, would look in terms of each.
a. According to the logical interpretation, probability or degree of confirmation is a logical relation between evidence and hypothesis. Degree of confirmation statements are analytic. There is no analogue of the rule of detachment (modus ponens), so inductive logic does not provide any basis for asserting inductive conclusions. There are certain methodological rules for the application of inductive logic, but these are not part of inductive logic itself. One such rule is the requirement of total evidence. For any given explanandum, taken as hypothesis, there are many true degree of confirmation statements corresponding to different possible evidence statements. The probability value to be chosen for practical applications is that given by the degree of confirmation statement embodying the total available (relevant) evidence. This is the value to be taken as a fair betting quotient, as an estimate of the relative frequency, or as the value to be employed in calculating the estimated utility.

I have long been seriously puzzled as to how Carnap's conception of logical probability could be consistently combined with Hempel's views on the nature of explanation. \(^{28}\) In the first place, for purposes of either deductive or inductive explanation, it is not clear how the universal or statistical generalizations demanded by Hempel's schema are to become available, for inductive logic has no means of providing such generalizations as detachable conclusions of any argument. In the second place, degree of confirmation statements embodying particular hypothesis statements and particular evidence statements are generally available, so inductive "inferences" from particulars to particulars are quite possible. In view of this fact, it is hard to see why inductive explanation should require any sort of general premise. In the third place, the relation between explanans and explanandum would seem to be a degree of confirmation statement, and not an argument with premises and conclusions in the usual sense. This last point seems to me to be profoundly important. Carnap's recognition that much, if not all, of inductive logic can proceed without inductive arguments—that is, without establishing conclusions on the basis of premises—is a deep insight that must be reckoned with. As a consequence, we must seriously question whether an explanation is an argument at all. This possibility arises, as I shall try to show later, even if one does not adhere to a logical interpretation of probability. Indeed, within an account of explanation in terms of the frequency interpretation, I shall argue contra Hempel that explanations are not arguments. \(^{29}\)

For the logical interpretation of probability, it would seem, an explana-

b. According to the new subjective or personalistic interpretation, probability is simply orderly opinion. \(^{30}\) The required orderliness is provided by the mathematical calculus of probability. One of the main concerns of the personalists is with the revision of opinion or degree of belief in the light of new evidence. The account of explanation suggested above fits very easily with the personalistic interpretation. At some time before the explanandum event occurs, the personalist would say, an individual has a certain degree of belief that it will occur—reflected, perhaps, in the kind of odds he is willing to give or take in betting on its occurrence. According to the present view of explanation, a personalist should require that the explanatory facts be such as to increase the prior confidence in the occurrence of the explanandum event. Of course, this view deliberately introduces into its treatment of scientific explanation a large degree of subjectivity, but this is not necessarily a defect in either the account of probability or the account of explanation.

c. According to the frequency interpretation, a probability is the limit of the relative frequency of an attribute in an infinite sequence of
events. A probability cannot, therefore, literally be assigned to a single event. Since there are many occasions when we must associate probabilities with single events, and not only with large (literally, infinite) aggregates of events, a way must be found for making the probability concept applicable to single events. We wager on single events—a single toss of the dice, a single horse race, a single deal of the cards. The results of our practical planning and effort depend upon the outcomes of single occurrences, not infinite sequences of them. If probability is to be a guide of life, it must be meaningful to apply probability values to single events. This problem of the single case has traditionally been one of the difficulties confronting the frequency theory, and we shall have to examine in some detail the way in which it can be handled. With regard to the single case, the frequency interpretation is on an entirely different footing from the logical and subjective interpretations. Neither of these latter interpretations faces any special problem of the single case, for statements about single events are perfectly admissible hypotheses for the logical theory, and the subjective theory deals directly with degrees of belief regarding single events.

The central topic of concern in this paper is the explanation of single events. If the frequency theory is to approach this problem at all, it must deal directly with the problem of the single case. To some, the fact that the frequency interpretation is faced with this special problem of the single case may constitute a compelling reason for concluding that the frequency interpretation is totally unsuitable for handling the explanation of single events. Such a conclusion would, I think, be premature. On the contrary, I would argue that a careful examination of the way in which the frequency interpretation handles the single case should prove extremely illuminating with respect to the general problem of inductive explanation. Although I have strong prejudices in favor of the frequency interpretation, this paper is, nevertheless, not the place to argue for them. In this context I would claim instead that frequencies play such an important role in any theory of probability that an examination of the problem of the single case cannot fail to cast light on the problem of explanation regardless of one's persuasion concerning interpretations of the probability concept.

d. In recent years there has been a good deal of discussion of the propensity interpretation of probability. This interpretation is so similar to the frequency interpretation in fundamental respects that everything I shall say about the problem of the single case for the frequency interpretation can be made directly applicable to the propensity interpretation.

4. The Single Case

Let A be an unending sequence of draws of balls from an urn, and let B be the class of red things. A is known as the reference class, and B the attribute class. The probability of red draws from this urn, P(A,B), is the limit of the relative frequency with which members of the reference class belong to the attribute class, that is, the limit of the relative frequency with which draws from the urn result in a red ball as the number of draws increases without any bound.

Frequentists like John Venn and Hans Reichenbach have dealt with the problem of the single case by assigning each single event to a reference class and by transferring the probability value from that reference class to the single event in question. Thus, if the limit of the relative frequency of red among draws from our urn is one-third, then we say that the probability of getting red on the next draw is one-third. In this way the meaning of the probability concept has been extended so that it applies to single events as well as to large aggregates.

The fundamental difficulty arises because a given event can be referred to any of a large number of reference classes, and the probability of the attribute in question may vary considerably from one of these to another. For instance, we could place two urns on a table, the one on the left containing only red balls, the one on the right containing equal numbers of red, white, and blue balls. The reference class A might consist of blind drawings from the right-hand urn, the ball being replaced and the urn thoroughly shaken after each draw. Another reference class A' might consist of drawings made alternately from the left- and right-hand urns. Infinitely many other reference classes are easily devised to which the next draw—the draw with which we are concerned—belongs. From which reference class shall we transfer our probability value to this single case? A method must be established for choosing the appropriate reference class. Notice, however, that there is no difficulty in selecting an attribute class. The question we ask determines the attribute class. We want to know the probability of getting red, so there is no further problem about the attribute class.
Reichenbach recommends adopting as a reference class "the narrowest class for which reliable statistics can be compiled." This principle is, as Reichenbach himself has observed, rather ambiguous. Since increasing the reliability of statistics generally tends to broaden the class and since narrowing the class often tends to reduce the reliability of the statistics, the principle involves two desiderata which pull in opposite directions. It seems that we are being directed to maximize two variables that cannot simultaneously be maximized. This attempt to extend the meaning of the probability concept to single cases fails to provide a method for associating a unique probability value with a given single event. Fully aware of this fact, Reichenbach insisted that the probability concept applies literally only to sequences; talk about the probability of a single event is "elliptical" and the extended meaning is "fictitious." The choice of a reference class, he maintained, is often dictated by practical rather than theoretical considerations.

Although Reichenbach has not said so, it seems reasonable to suppose that he was making a distinction similar to that made by Carnap between the principles belonging to inductive logic and methodological rules for the application of inductive logic. The requirement of total evidence, it will be recalled, is a methodological rule for the application of inductive logic. Reichenbach could be interpreted as suggesting analogously that probability theory itself is concerned only with limit statements about relative frequencies in infinite sequences of events, whereas the principle for selection of a reference class stands as a methodological rule for the practical application of probability statements. (A much stronger analogy between the requirement of total evidence and the principle for choosing a reference class for a single case will be shown below.) In fact, for Reichenbach, it might have been wise to withhold the term "probability" from single events, reserving his term "weight" for this purpose. We could then say that practical considerations determine what probability should be chosen to serve as a weight for a particular single event. The relative practical importance of reliability and precision would then determine the extent to which narrowness gives way to reliability of statistics (or conversely) in determining the appropriate reference class.

Although Reichenbach's formulation of the principle for the selection of reference classes is not entirely satisfactory, his intention seems fairly clear. In order to transfer a probability value from a sequence to a single case, it is necessary to have some basis for ascertaining the probability in that sequence. The reference class must, therefore, be broad enough to provide the required number of instances for examination to constitute evidence for an inductive inference. At the same time, we want to avoid choosing a reference class so broad that it includes cases irrelevant to the one with which we are concerned.

Statistical relevance is the essential notion here. It is desirable to narrow the reference class in statistically relevant ways, but not in statistically irrelevant ways. When we choose a reference class to which to refer a given single case, we must ask whether there is any statistically relevant way to subdivide that class. If so, we may choose the narrower subclass that results from the subdivision; if no statistically relevant way is known, we must avoid making the reference class any narrower. Consider, for example, the probability that a particular individual, John Smith, will still be alive ten years hence. To determine this probability, we take account of his age, sex, occupation, and health; we ignore his eye color, his automobile license number, and his last initial. We expect the relative frequency of survival for ten more years to vary among the following reference classes: humans, Americans, American males, forty-two-year-old American males, forty-two-year-old American male steeplejacks, and forty-two-year-old American male steeplejacks suffering from advanced cases of lung cancer. We believe that the relative frequency of survival for another ten years is the same in the following classes: forty-two-year-old American male steeplejacks with advanced cases of lung cancer, forty-two-year-old blue-eyed American male steeplejacks with advanced cases of lung cancer, and forty-two-year-old blue-eyed American male steeplejacks with even automobile license plate numbers who suffer from advanced cases of lung cancer.

Suppose we are dealing with some particular object or event \( x \), and we seek to determine the probability (weight) that it has attribute \( B \). Let \( x \) be assigned to a reference class \( A \), of which it is a member. \( P(A,B) \) is the probability of this attribute within this reference class. A set of mutually exclusive and exhaustive subclasses of a class is a partition of that class. We shall often be concerned with partitions of reference classes into two subclasses; such partitions can be effected by a property \( C \) which divides the class \( A \) into two subclasses, \( A.C \) and \( A.C \). A property \( C \) is said to be statistically relevant to \( B \) within \( A \) if and only if \( P(A,C,B) \neq P(A,B) \). This notion of statistical relevance is the fundamental concept upon which I hope to build an explication of inductive explanation.

In his development of a frequency theory based essentially upon the concept of randomness, Richard von Mises introduced the notion of a place selection: "By a place selection we mean the selection of a partial
A place selection effects a partition of a reference class into two subclasses, elements of the place selection and elements not included in the place selection. In the reference class of draws from our urn, every third draw starting with the second, every kth draw where k is prime, every draw following a red result, every draw made with the left hand, and every draw made while the sky is cloudy all would be place selections. “Every draw of a red ball” and “every draw of a ball whose color is at the opposite end of the spectrum from violet” do not define place selections, for membership in these classes cannot be determined without reference to the attribute in question.

A place selection may or may not be statistically relevant to a given attribute in a given reference class. If the place selection is statistically irrelevant to an attribute within a reference class, the probability of that attribute within the subclass determined by the place selection is equal to the probability of that attribute within the entire original reference class. If every place selection is irrelevant to a given attribute in a given sequence, von Mises called the sequence random. If every property that determines a place selection is statistically irrelevant to B in A, I shall say that A is a homogeneous reference class for B. A reference class is homogeneous if there is no way, even in principle, to effect a statistically relevant partition without already knowing which elements have the attribute in question and which do not. Roughly speaking, each member of a homogeneous reference class is a random member.

The aim in selecting a reference class to which to assign a single case is not to select the narrowest, but the widest, available class. However, the reference class should be homogeneous, and achieving homogeneity requires making the reference class narrower if it was not already homogeneous. I would reformulate Reichenbach’s method of selection of a reference class as follows: choose the broadest homogeneous reference class to which the single event belongs. I shall call this the reference class rule.

Let me make it clear immediately that, although I regard the above formulation as an improvement over Reichenbach’s, I do not suppose that it removes all ambiguities about the selection of reference classes either in principle or in practice. In principle it is possible for an event to belong to two equally wide homogeneous reference classes, and the probabilities of the attribute in these two classes need not be the same. For instance, suppose that the drawing from the urn is not random and that the limit of the relative frequency of red for every kth draw (k prime) is 1/4, whereas the limit of the relative frequency of red for every even draw is 3/4. Each of these subsequences may be perfectly random; each of the foregoing place selections may, therefore, determine a homogeneous reference class. Since the intersection of these two place selections is finite, it does not determine a reference class for a probability. The second draw, however, belongs to both place selections; in this fictitious case there is a genuine ambiguity concerning the probability to be taken as the weight of red on the second draw.

In practice we often lack full knowledge of the properties relevant to a given attribute, so we do not know whether our reference class is homogeneous or not. Sometimes we have strong reason to believe that our reference class is not homogeneous, but we do not know what property will affect a statistically relevant partition. For instance, we may believe that there are causal factors that determine which streptococcus infections will respond to penicillin and which ones will not, but we may not yet know what these causal factors are. When we know or suspect that a reference class is not homogeneous, but we do not know how to make any statistically relevant partition, we may say that the reference class is epistemically homogeneous. In other cases, we know that a reference class is inhomogeneous and we know what attributes would effect a statistically relevant partition, but it is too much trouble to find out which elements belong to each subclass of the partition. For instance, we believe that a sufficiently detailed knowledge of the initial conditions under which a coin is tossed would enable us to predict (perfectly or very reliably) whether the outcome will be heads or tails, but practically speaking we are in no position to determine these initial conditions or make the elaborate calculations required to predict the outcome. In such cases we may say that the reference class is practically homogeneous.

The reference class rule remains, then, a methodological rule for the application of probability knowledge to single events. In practice we attempt to refer our single cases to classes that are practically or epistemically homogeneous. When something important is at stake, we may try to extend our knowledge in order to improve the degree of homogeneity we can achieve. Strictly speaking, we cannot meaningfully refer to degrees of homogeneity until a quantitative concept of homogeneity has been provided. That will be done in section 6 below.

It would, of course, be a serious methodological error to assign a single case to an inhomogeneous reference class if neither epistemic nor practical
considerations prevent partitioning to achieve homogeneity. This fact constitutes another basis for regarding the reference class rule as the counterpart of the requirement of total evidence. The requirement of total evidence demands that we use all available relevant evidence; the reference class rule demands that we partition whenever we have available a statistically relevant place selection by means of which to effect the partition.

Although we require homogeneity, we must also prohibit partitioning of the reference class by means of statistically irrelevant place selections. The reason is obvious. Irrelevant partitioning reduces, for no good reason, the inductive evidence available for ascertaining the limiting frequency of our attribute in a reference class that is as homogeneous as we can make it. Another important reason for prohibiting irrelevant partitions will emerge below when we discuss the importance of multiple homogeneous reference classes.

A couple of fairly obvious facts about homogeneous reference classes should be noted at this point. If all A’s are B, A is a homogeneous reference class for B. (Somewhat counterintuitively, perhaps, B occurs perfectly randomly in A.) In this case, \( P(A, B) = 1 \) and \( P(A, C, B) = 1 \) for any C whatever; consequently, no place selection can yield a probability for B different from that in the reference class A. Analogously, A is homogeneous for B if no A’s are B. In the frequency interpretation, of course, \( P(A, B) \) can equal one even though not all A’s are B. It follows that a probability of one does not entail that the reference class is homogeneous.

Some people maintain, often on a priori grounds, that A is homogeneous (not merely practically or epistemically homogeneous) for B only if all A’s are B or no A’s are B; such people are determinists. They hold that causal factors always determine which A’s are B and which A’s are not B; these causal factors can, in principle, be discovered and used to construct a place selection for making a statistically relevant partition of A. I do not believe in this particular form of determinism. It seems to me that there are cases in which A is a homogeneous reference class for B even though not all A’s are B. In a sample of radioactive material a certain percentage of atoms disintegrate in a given length of time; no place selection can give us a partition of the atoms for which the frequency of disintegration differs from that in the whole sample. A beam of electrons is shot at a potential barrier and some pass through while others are reflected; no place selection will enable us to make a statistically relevant partition in the class of electrons in the beam. A beam of silver atoms is sent through a strongly inhomogeneous magnetic field (Stern-Gerlach experiment); some atoms are deflected upward and some are deflected downward, but there is no way of partitioning the beam in a statistically relevant manner. Some theorists maintain, of course, that further investigation will yield information that will enable us to make statistically relevant partitions in these cases, but this is, at present, no more than a declaration of faith in determinism. Whatever the final resolution of this controversy, the homogeneity of A for B does not logically entail that all A’s are B. The truth or falsity of determinism cannot be settled a priori.

The purpose of the foregoing excursus on the frequency treatment of the problem of the single case has been to set the stage for a discussion of the explanation of particular events. Let us reconsider some of our examples in the light of this theory. The relative frequency with which we encounter instances of water-soluble substances in the normal course of things is noticeably less than one; therefore, the probability of water solubility in the reference class of samples of unspecified substances is significantly less than one. If we ask why a particular sample of unspecified material has dissolved in water, the prior weight of this explanation of its occurrence is less than one as referred to the class of samples of unspecified substances. This broad reference class is obviously inhomogeneous with respect to water solubility. If we partition it into the subclass of samples of table salt and samples of substances other than table salt, it turns out that every member of the former subclass is water-soluble. The reference class of samples of table salt is homogeneous with respect to water solubility. The weight for the single case, referred to this homogeneous reference class, is much greater than its prior weight. By referring the explanandum event to a homogeneous reference class and substantially increasing its weight, we have provided an inductive explanation of its occurrence. As the discussion develops, we shall see that the homogeneity of the reference class is the key to the explanation. The increase in weight is a fortunate dividend in many cases.

If we begin with the reference class of samples of table salt, asking why this sample of table salt dissolves in water, we already have a homogeneous reference class. If, however, we subdivide that reference class into hexed and unhxed samples, we have added nothing to the explanation of dissolving, for no new probability value results and we have not made the already homogeneous reference class any more homo-
The original reference class of samples of unspecified substances can be partitioned into hexed and unhxed samples. If this partition is accomplished by means of a place selection—that is, if the hexing is done without reference to previous knowledge about solubility—the probabilities of water solubility in the subclasses will be no different from the probability in the original reference class. The reference class of hexed samples of unspecified substances is no more homogeneous than the reference class of samples of unspecified substances; moreover, it is narrower. The casting of a dissolving spell is statistically irrelevant to water solubility, so it cannot contribute to the homogeneity of the reference class, and it must not be used in assigning a weight to the single case. For this reason it contributes nothing to the explanation of the fact that this substance dissolves in water.

The vitamin C example involves the same sort of consideration. In the class of colds in general, there is a rather high frequency of recovery within a week. In the narrower reference class of colds for which the victim has taken vitamin C, the frequency of recovery within a week is no different. Vitamin C is not efficacious, and that fact is reflected in the statistical irrelevance of administration of vitamin C to recovery from a cold within a week. Subdivision of the reference class in terms of administration of vitamin C does not yield a more homogeneous reference class and, consequently, does not yield a higher weight for the explanandum event. In similar fashion, we know that noisemaking and shooting off fireworks are statistically irrelevant to the reappearance of the moon after an eclipse, and we know that our friend’s loud moaning is statistically irrelevant to finding a wild tiger in Times Square. In none of these horrible examples do the “explanatory facts” contribute anything toward achieving a homogeneous reference class or to an increase of posterior weight over prior weight.

5. The Requirement of Maximal Specificity

In his 1962 essay, “Deductive-Nomological vs. Statistical Explanation,” Hempel takes note of what he calls “the ambiguity of statistical systematization,” and he invokes a requirement of total evidence to overcome it. This kind of ambiguity is seen to arise out of the possibility of referring a given event to several different reference classes, and Hempel explicitly compares his requirement of total evidence to Reichenbach’s rule of selecting the narrowest reference class for which reliable statistics are available.

These requirements do not exclude any of the counterexamples introduced above, in particular, the “explanations” of John Jones’s recovery from his cold (5) and of his remission of neurotic symptoms (6).

In the 1965 essay, “Aspects of Scientific Explanation,” Hempel distinguishes the ambiguity of statistical explanation from the epistemic ambiguity of statistical explanation. The former ambiguity arises from the existence of different reference classes, with different relative frequencies of the attribute in question, to which a given single case may be referred. The latter ambiguity arises from the occurrence of items in our body of accepted scientific knowledge which lead us to refer the given single case to different reference classes. In order to cope with the problem of epistemic ambiguity, Hempel introduces the requirement of maximal specificity for inductive-statistical explanations as follows:

Consider a proposed explanation of the basic statistical form

\[ p(G,F_s) = r_s \]

Let \( F \) be the conjunction of the premises, and, if \( K \) is the set of all statements accepted at the given time, let \( k \) be a sentence that is logically equivalent to \( K \) (in the sense that \( k \) is implied by \( K \) and in turn implies every sentence in \( K \)). Then, to be rationally acceptable in the knowledge situation represented by \( K \), the proposed explanation (30) must meet the following condition (the requirement of maximal specificity): If \( s,k \) implies that \( b \) belongs to a class \( F_s \), and that \( F_s \) is a subclass of \( F \), then \( s,k \) must also imply a sentence specifying the statistical probability of \( G \) in \( F_s \), say

\[ p(G,F_s) = r_s \]

Here, \( r_s \) must equal \( r \) unless the probability statement just cited is simply a theorem of mathematical probability theory.

Hempel goes on to remark that

the requirement of maximal specificity, then, is here tentatively put forward as characterizing the extent to which the requirement of total evidence properly applies to inductive-statistical explanations. The general idea thus suggested comes to this: In formulating or appraising I-S explanation, we should take into account all that information provided by \( K \) which is of potential explanatory relevance to the explanandum event; i.e., all pertinent statistical laws, and such particular facts as might be connected, by the statistical laws, with the explanandum event.

There are two immediate objections to the requirement of maximal specificity as thus formulated. First, like the requirement of total evi-
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dence, it fails to rule out such counterexamples as (5), the explanation of John Jones's recovery from his cold on account of taking vitamin C. This "explanation" is not blocked by the requirement, for, as we have noted, there is no narrower class for which the probability of recovery from a cold would be different. The trouble is that the class invoked in the "explanation" is too narrow; the class of people with colds—regardless of medication—already satisfies the requirement. The difficulty seems to arise from the emphasis upon narrowness in Reichenbach's formulation of his reference class rule. As I reformulate the rule, we seek the broadest homogeneous reference class. However, this problem is not serious, for it can be circumvented by a minor reformulation of Hempel's requirement. We need simply to add to Hempel's formulation the condition that \( F \) is the largest class that satisfies it. This addition would result in what might be called the requirement of the maximal class of maximal specificity.

Second, the final sentence in the requirement seems designed to achieve epistemic homogeneity, but it is not strong enough to do so. In explaining the reason for the qualification "\( r \) must equal \( r \) unless the probability statement just cited is simply a theorem of mathematical probability theory," Hempel points out that \( F \cap G \) is always a subset of \( F \), but obviously \( p(G,F) = 1 \) by the mathematical calculus alone. Hence, without the qualification of the "unless" clause, the requirement of maximal specificity could never be nontrivially fulfilled.

When von Mises attempted to characterize randomness in precisely the sense that is at issue here, he pointed out that the relative frequency in a subsequence must be invariant if the subsequence is determined by a place selection, but that it need not be invariant for subsequences determined by other types of selections. For instance, let \( F \) be the class of draws from a particular urn and \( G \) the drawing of a red ball. Suppose, as a matter of fact, that the person drawing from the urn sometimes draws with his left hand and sometimes with his right, and that every draw with the left hand produces a red ball, whereas some draws with the right produce balls of other colors. Let \( H \) be the class of draws with the left hand. Under these conditions \( F \) does not fulfill the requirement of maximal specificity, for \( F \cap H \) is a subclass of \( F \) in which \( G \) has a different probability than it does in \( F \). Obviously, the fact that all \( H \)'s are \( G \) is not a theorem of the probability calculus; it is a physical fact.

Now, to take a different example for purposes of comparison, suppose that everything is the same as in the foregoing case, except that \( p(G,F) = p(G,H,F) \)—that is, the frequency with which red appears is the same regardless of which hand is used for the draw, and, so far as \( H \) is concerned, \( F \) satisfies the requirement of maximal specificity. Let \( J \) be the class of draws of a ball whose color is at the opposite end of the visible spectrum from violet. Clearly, all \( J \)'s are \( G \), but I consider this to be an extremely highly confirmed matter of fact, not a theorem of the probability calculus or of pure logic. It follows immediately that \( p(G,F) \neq p(G,J,F) = 1 \). Hence, by Hempel's formulation \( F \) still violates the requirement of maximal specificity, and so obviously will any class \( F \) except when either all \( F \)'s are \( G \) or no \( F \)'s are \( G \). In the first case we wanted to say that the class \( F \) violates maximal specificity because of the empirical generalization that all \( F \)'s are \( G \); in the second case we were forced unwillingly to admit that \( F \) violates maximal specificity because of the empirical generalization that all \( F \)'s are \( G \). The difference is, roughly speaking, that we can know whether a draw is being made with the left hand before we see the result, but we cannot know that a draw results in a color at the opposite end of the visible spectrum from violet until we know what the result is. To characterize this difference precisely might be difficult, but it is required for von Mises's concept of randomness, for my concept of homogeneity, and, I believe, for a satisfactory reformulation of Hempel's requirement of maximal specificity.

When the foregoing two difficulties are overcome, I believe that Hempel's requirement of maximal specificity will become equivalent to my concept of epistemic homogeneity, and, I believe, it will rule out counterexamples of the sort that I have presented above. It will rule out the deductive examples as well as the inductive examples if we construe the deductive examples as limiting cases of inductive explanation and demand that they also satisfy the requirement of the maximal class of maximal specificity. If, however, these two rather straightforward revisions of Hempel's requirement of maximal specificity were all that is needed to fix up all of the difficulties in Hempel's account of explanation of particular events, it would have been an outrageous imposition upon the good humor of the reader to have taken so many pages to say so. The fact is quite otherwise. These rather lengthy discussions have been intended to lay the foundations for further considerations that will lead to far-reaching differences with Hempel's basic conceptions. The point that we have arrived at so far is that requirements (i) through (iv) above (see page 176) still stand, except that (iv) now contains the corrected requirement of the maximal class of maximal specificity. Requirement (i) still stands, and it demands for inductive explanation that the explanandum be made highly probable (to a degree \( \geq \) some chosen number \( r \)). According to this conception, an explanation is still regarded
as an argument to the effect that a certain event was to be expected by virtue of certain explanatory facts. I shall maintain that we have still failed to come to grips adequately with the problem of relevance of the explanans to the explanandum.

6. Prior Weights and Degree of Inhomogeneity

One of the fundamental features of explanation according to the approach I have been suggesting is that it concerns the relations between the prior weight and the posterior weight of the explanandum event. I have attempted to indicate how a frequency theorist may get weights from probabilities, by application of the reference class rule, but so far nothing has been said about the prior probability from which the prior weight is to be taken. Let us now consider that question.

If we look at Hempel's schemata (3) and (4) for deductive and inductive explanation, respectively, we see that the form of the conclusion in each case is "A is C." However, the explanatory question to which the proffered explanation attempts to supply an answer has a more complex form. We do not ask, "Why is this thing yellow?" We ask, "Why is this Bunse flame yellow?" We do not ask, "Why does this thing disappear?" We ask "Why does this streptococcus infection disappear?" In every case, I think, the question takes the form, "Why is this A which is also B?" The answer then takes the form, "Because this x is also C." C must be an attribute that is statistically relevant to B within the reference class A.

The explanatory question, it seems to me, furnishes an original reference class A to which the explanandum event is referred. The probability of the attribute B within that reference class A is the prior weight for purposes of that explanation; normally, the class A would be epistemically homogeneous prior to the furnishing of the explanation. In cases of genuine explanation, the reference class A is not actually homogeneous with respect to the attribute B, and so a further property C is invoked to achieve a homogeneous (or more nearly homogeneous) reference class A.C for purposes of establishing a posterior weight. If the original reference class A is homogeneous, the introduction of an irrelevant characteristic C to narrow the reference class is not only without explanatory value but is actually methodologically detrimental. In a later section of this paper, I shall offer a more formal characterization of such explanations, showing how the prior weight and posterior weight figure in the schema.

We can now introduce a quantitative measure of the degree of inhomogeneity of the reference class A which will, in a sense, indicate what has been achieved when that class is partitioned into homogeneous subclasses. Let the sequence $x_1, x_2, \ldots, x_n, \ldots$ be the ordered set A whose degree of inhomogeneity with respect to the attribute B is to be measured. Assume that there exists a unique homogeneous partition $C_1, \ldots, C_k$ of A with respect to B. Let $P(A,B) = p$, and let $P(A,C_i,B) = p_i$. Assign to each element $x_i$ of A the weight $w_i = p_i$ associated with that compartment of the partition $C_i$ to which $x_i$ belongs, that is, if $x_i \in A.C_i$, then $w_i = p_i = P(A,C_i,B)$. Evidently, $p = w$ is the prior weight of each $x_i$ in A with respect to the attribute B, whereas $w_i$ is its posterior weight in the homogeneous partition of A.

If we say, in effect, that $w_i$ is the correct weight to assign to $x_i$, then we may say that $p - w_i$ is the "error" that would result from the use of the prior weight instead of the posterior weight. Squaring to make this quantity nonnegative, we may take the squared error $(p - w_i)^2$ as a measure of error. Then

$$
\sum_{i=1}^{n} (p - w_i)^2
$$

is the cumulative squared error involved in using the prior weight on the first $n$ elements of A, and

$$
\frac{1}{n} \sum_{i=1}^{n} (p - w_i)^2
$$

is the mean squared error. The limit of the mean squared error as $n$ increases without bound will be taken to represent the inhomogeneity of the reference class A with respect to the attribute B, but for reasons to be mentioned in a moment, I shall multiply this number by a constant for purposes of defining degree of inhomogeneity.

If we think of the weights $w_i$ simply as numbers that exhibit a certain frequency distribution, then it is clear that $p$ represents their mean value, that is,

$$
\lim_{n \to \infty} \frac{1}{n} \sum_{i=1}^{n} w_i = p.
$$

Then

$$
\lim_{n \to \infty} \frac{1}{n} \sum_{i=1}^{n} (p - w_i)^2 = \sigma^2
$$
is a measure of the dispersion of the $w_i$ known as the variance (which is the square of the standard deviation $\sigma$).

If the reference class $A$ is already homogeneous, the error is identically zero for every $i$, and the variance is zero. The maximum degree of inhomogeneity seems to be represented by the case in which $B$ occurs with probability $1/2$ in $A$, but there is a partition of $A$ such that every element of $C_1$ is $B$ and no element of $C_2$ is $B$. These conditions would obtain, for instance, if a sequence of coin tosses consisted of tosses of a two-headed coin randomly interspersed with tosses of a two-tailed coin, each of the two being used with equal frequency. In this case $p = 1/2$, and each $w_i$ is either $0$ or $1$. Thus, the error on each toss is $1/2$, and the mean squared error (for each $n$) as well as the variance is $1/4$. In order to have a convenient measure of degree of homogeneity, let us multiply the variance by $4$, so that degree of inhomogeneity ranges from $0$ to $1$. We can then adopt the following formal definition:

The degree of inhomogeneity of class $A$ with respect to attribute $B = 4\sigma^2$.

It is worth noting that the transition from an inhomogeneous reference class $A$ to a set of homogeneous subclasses $A,C_i$ provides an increase of information. In the extreme example given above, when we have only the original inhomogeneous reference class of coin tosses, we know only that the probability of a head is $1/2$, and we know nothing about whether an individual toss will result in a head or a tail. When the class is partitioned into two subclasses, one consisting of tosses of the two-headed coin and the other consisting of tosses with the two-tailed coin, we have complete information regarding the outcome of each toss; each result can be predicted with certainty. 

7. Causal and Statistical Relevance

The attempt to explicate explanation in terms of probability, statistical relevance, and homogeneity is almost certain to give rise to a standard objection. Consider the barometer example introduced by Michael Scriven in a discussion of the thesis of symmetry between explanation and prediction—a thesis whose discussion I shall postpone until a later section. If the barometer in my house shows a sudden drop, a storm may be predicted with high reliability. But the barometric reading is only an indicator; it does not cause the storm and, according to Scriven, it does not explain its occurrence. The storm is caused by certain widespread atmospheric conditions, and the behavior of the barometer is merely symptomatic of them. "In explanation we are looking for a cause, an event that not only occurred earlier but stands in a special relation to the other event. Roughly speaking, the prediction requires only a correlation, the explanation more."  

The objection takes the following form. There is a correlation between the behavior of the barometer and the occurrence of storms. If we take the general reference class of days in the vicinity of my house and ask for the probability of a storm, we get a rather low prior probability. If we partition that reference class into two subclasses, namely, days on which there is a sudden drop in the barometer and days on which there is not, we have a posterior probability of a storm in the former class much higher than the prior probability. The new reference class is far more homogeneous than the old one. Thus, according to the view I am suggesting, the drop in barometric reading would seem to explain the storm.

I am willing to admit that symptomatic explanations seem to have genuine explanatory value in the absence of knowledge of causal relations, that is, as long as we do not know that we are dealing only with symptoms. Causal explanations supersede symptomatic ones when they can be given, and when we suspect we are dealing with symptoms, we look hard for a causal explanation. The reason is that a causal explanation provides a more homogeneous reference class than does a symptomatic explanation. Causal proximity increases homogeneity. The reference class of days on which there is a local drop in barometric pressure inside my house, for instance, is more homogeneous than the reference class of days on which my barometer shows a sudden drop, for my barometer may be malfunctioning. Similarly, the reference class of days on which there is a widespread sudden drop in atmospheric pressure is more homogeneous than the days on which there is a local drop, for the house may be tightly sealed or the graduate students may be playing a joke on me. It is not that we obtain a large increase in the probability of a storm as we move from one of these reference classes to another; rather, each progressively better partitioning makes the preceding partitioning statistically irrelevant.

It will be recalled that the property $C$ is statistically irrelevant to the attribute $B$ in the reference class $A$ if $P(A,B) = P(C,A,B)$. The probability of a storm on a day when there is a sudden drop in atmospheric pressure and when my barometer executes a sudden drop is precisely the same as the probability of a storm on a day when there is a sudden
widespread drop in atmospheric pressure. To borrow a useful notion from Reichenbach, we may say that the sudden widespread drop in atmospheric pressure screens off the drop in barometer reading from the occurrence of the storm. The converse relation does not hold. The probability of a storm on a day when the reading on my barometer makes a sudden drop is not equal to the probability of a storm on a day when the reading on my barometer makes a sudden drop and there is a widespread drop in the atmospheric pressure. The sudden drop in barometric reading does not screen off the sudden widespread drop in atmospheric pressure from the occurrence of the storm.

More formally, we may say that $D$ screens off $C$ from $B$ in reference class $A$ iff (if and only if)

$$P(A,C,D,B) = P(A,D,B) \neq P(A,C,B).$$

For purposes of the foregoing example, let $A =$ the class of days in the vicinity of my house, let $B =$ the class of days on which there is an occurrence of a storm, let $C =$ the class of days on which there is a sudden drop in reading on my barometer, and let $D =$ the class of days on which there is a widespread drop in atmospheric pressure in the area in which my house is located. By means of this formal definition, we see that $D$ screens off $C$ from $B$, but $C$ does not screen off $D$ from $B$. The screening-off relation is, therefore, not symmetrical, although the relation of statistical relevance is symmetrical.

When one property in terms of which a statistically relevant partition in a reference class can be effected screens off another property in terms of which another statistically relevant partition of that same reference class can be effected, then the screened-off property must give way to the property which screens it off. This is the screening-off rule. The screened-off property then becomes irrelevant and no longer has explanatory value. This consideration shows how we can handle the barometer example and a host of others, such as the explanation of measles in terms of spots, in terms of exposure to someone who has the disease, and in terms of the presence of the virus. The unwanted "symptomatic explanations" can be blocked by use of the screening-off concept, which is defined in terms of statistical irrelevance alone. We have not found it necessary to introduce an independent concept of causal relation in order to handle this problem. Reichenbach believed it was possible to define causal relatedness in terms of screening-off relations; but whether his program can be carried through or not, it seems that many causal relations exhibit the desired screening-off relations.

8. Explanations with Low Weight

According to Hempel, the basic requirement for an inductive explanation is that the posterior weight (as I have been describing it) must be high, whereas I have been suggesting that the important characteristic is the increase of the posterior weight over the prior weight as a result of incorporating the event into a homogeneous reference class. The examples discussed thus far satisfy both of these desiderata, so they do not serve well to discriminate between the two views. I would maintain, however, that when the prior weight of an event is very low, it is not necessary that its posterior weight be made high in order to have an inductive explanation. This point is illustrated by the well-known paresis example.

No one ever contracts paresis unless he has had latent syphilis which has gone untreated, but only a small percentage of victims of untreated latent syphilis develop paresis. Still, it has been claimed, the occurrence of paresis is explained by the fact that the individual has had syphilis. This example has been hotly debated because of its pertinence to the issue of symmetry between explanation and prediction—an issue I still wish to postpone. Nevertheless, the following observations are in order. The prior probability of a person contracting paresis is very low, and the reference class of people in general is inhomogeneous. We can make a statistically relevant partition into people who have untreated latent syphilis and those who do not. (Note that latent syphilis screens off primary syphilis and secondary syphilis.) The probability that a person with untreated latent syphilis will contract paresis is still low, but it is considerably higher than the prior probability of paresis among people in general. To cite untreated latent syphilis as an explanation of paresis is correct, for it does provide a partition of the general reference class which yields a more homogeneous reference class and a higher posterior weight for the explanandum event.

When the posterior weight of an event is low, it is tempting to think that we have not fully explained it. We are apt to feel that we have not yet found a completely homogeneous reference class. If only we had fuller understanding, we often believe, we could sort out causal antecedents in order to be able to say which cases of untreated latent syphilis will become paretic and which ones will not. With this knowledge we would be able to partition the reference class of victims of untreated latent syphilis into two subclasses in terms of these causal antecedents so that all (or an overwhelming majority of) members of one subclass will
develop paresis whereas none (or very few) of the members of the other will become paretic. This conviction may be solidly based upon experience with medical explanation, and it may provide a sound empirical basis for the search for additional explanatory facts—more relevant properties in terms of which to improve the homogeneity of the reference class. Nevertheless, the reference class of untreated latent syphilics is (as I understand it) epistemically homogeneous in terms of our present knowledge, so we have provided the most adequate explanation possible in view of the knowledge we possess. 86

A parallel example could be constructed in physics where we have much greater confidence in the actual homogeneity of the reference class. Suppose we had a metallic substance in which one of the atoms experienced radioactive decay within a particular, small time period, say one minute. For purposes of the example, let us suppose that only one such decay occurred within that time period. When asked why that particular atom decayed, we might reply that the substance is actually an alloy of two metals, one radioactive (for example, uranium 238) and one stable (for example, lead 206). Since the half-life of U238 is $4.5 \times 10^9$ years, the probability of a given uranium atom’s decaying in an interval of one minute is not large, yet there is explanatory relevance in pointing out that the atom that did decay was a U238 atom. 87 According to the best theoretical knowledge now available, the class of U238 atoms is homogeneous with respect to radioactive decay, and there is in principle no further relevant partition that can be made. Thus, it is not necessary to suppose that examples such as the paresis case derive their explanatory value solely from the conviction that they are partial explanations which can someday be made into full explanations by means of further knowledge. 88

There is one further way to maintain that explanations of improbable events involve, nevertheless, high probabilities. If an outcome is improbable (though not impossible) on a given trial, its occurrence at least once in a sufficiently large number of trials can be highly probable. For example, the probability of getting a double six on a toss of a pair of standard dice is $1/36$; in twenty-five tosses the probability of at least one double six is over one-half. No matter how small the probability $p$ of an event on a single trial, provided $p > 0$, and no matter how large $r$, provided $r < 1$, there is some $n$ such that the probability of at least one occurrence in $n$ trials is greater than $r$. 89 On this basis, it might be claimed, we explain improbable events by saying, in effect, that given enough trials the event is probable. This is a satisfactory explanation of the fact that certain types of events occur occasionally, but it still leaves open the question of how to explain the fact that this particular improbable event happened on this particular occasion.

To take a somewhat more dramatic example, each time an alpha particle bombards the potential barrier of a uranium nucleus, it has a chance of about $10^{-28}$ of tunneling through and escaping from the nucleus; one can appreciate the magnitude of this number by noting that whereas the alpha particle bombards the potential barrier about $10^{22}$ times per second, the half-life of uranium is of the order of a billion years. 90 For any given uranium atom, if we wait long enough, there is an overwhelmingly large probability that it will decay, but if we ask, “Why did the alpha particle tunnel out on this particular bombardment of the potential barrier?” the only answer is that in the homogeneous reference class of approaches to the barrier, it has a 1-in-$10^{28}$ chance of getting through. I do not regard the fact that it gets through on a particular trial inexplicable, but certainly anyone who takes explanations to be arguments showing that the event was to be expected must conclude that this fact defies all explanation.

9. Multiple Homogeneity

The paresis example illustrates another important methodological point. I have spoken so far as if one homogeneous reference class were sufficient for the explanation of a particular event. This, I think, is incorrect. When a general reference class is partitioned, we can meaningfully ask about the homogeneity of each subclass in the partition (as we did in defining degree of inhomogeneity above). To be sure, when we are attempting to provide an explanation of a particular explanandum event $x$, we focus primary attention upon the subclass to which $x$ belongs. Nevertheless, I think we properly raise the question of the homogeneity of other subclasses when we evaluate our explanation. In the paresis example we may be convinced that the reference class of untreated latent syphilis is inhomogeneous, but the complementary class is perfectly homogeneous. Since no individuals who do not have untreated latent syphilis develop paresis, no partition statistically relevant to the development of paresis can be made in the reference class of people who do not have untreated latent syphilis.

Consider the table salt example from the standpoint of the homogeneity of more than one reference class. Although the reference class of samples of table salt is completely homogeneous for water solubility, the complementary class certainly is not. It is possible to make further partitions in the class of samples of substances other than table salt.
which are statistically relevant to water solubility: samples of sand, wood, and gold are never water soluble; samples of baking soda, sugar, and rock salt always are.

If we explain the fact that this sample dissolves in water by observing that it is table salt and all table salt is water soluble, we may feel that the explanation is somewhat inadequate. Some theorists would say that it is an adequate explanation, but that we can equally legitimately ask for an explanation of the general fact that table salt is water soluble. Although I have great sympathy with the idea that general facts need explanation and are amenable to explanation, I think it is important to recognize the desideratum of homogeneity of the complementary reference class. I think we may rightly claim fully adequate explanation of a particular fact when (but not necessarily only when) the original reference class A, with respect to which its prior probability is assessed, can be partitioned into two subclasses A.C and A.C, each of which is homogeneous for the attribute in question. In the ideal case all A.C's are B and no A.C's are B—that is, if x is A, then x is B if and only if x is C. However, there is no reason to believe that Nature is so accommodating as to provide in all cases even the possibility in principle of such fully deterministic explanations.

We now have further reason to reject the dissolving spell explanation for the fact that a sample of table salt dissolves in water. If we partition the general reference class of samples of unspecified substances into hexed samples of table salt and all other samples of substances, this latter reference class is less homogeneous than the class of all samples of substances other than table salt, for we know that all unhexed samples of table salt are water soluble. To make a statistically irrelevant partition not only reduces the available statistics, it also reduces the homogeneity of the complementary reference class. This consideration also applies to such other examples as John Jones and his wife's birth control pills.

It would be a mistake to suppose that it must always be possible in principle to partition the original reference class A into two homogeneous subclasses. It may be that the best we can hope for is a partition into k subclasses A.C, each completely homogeneous, and such that P(A.C.B) \neq P(A.C.B) if i \neq j. This is the multiple homogeneity rule. It expresses the fundamental condition for adequate explanation of particular events, and it will serve as a basis for the general characterization of deductive and inductive explanation.

The multiple homogeneity requirement provides, I believe, a new way to attack a type of explanation that has been extremely recalcitrant in the face of approaches similar to Hempel's, namely, the so-called functional explanation. We are told, for example, that it is the function of hemoglobin to transport oxygen from the lungs to the cells in the various parts of the organism. This fact is taken to provide an explanation for the blood's hemoglobin content. Trouble arises when we try to fit such explanations to schemata like (3) for deductive explanation, for we always seem to end up with a necessary condition where we want a sufficient condition. Consider the following:

(11) Hemoglobin transports oxygen from the lungs to the other parts of the body (in an animal with a normally functioning circulatory system).
This animal has hemoglobin in its blood.
This animal has oxygen transported from its lungs to the other parts of its body.

Here we have a valid deductive argument, but unfortunately the explanandum is a premise and the conclusion is part of the explanans. This will never do. Let us interchange the second premise and the conclusion, so that the explanandum becomes the conclusion and the explanans consists of all of the premises, as follows:

(12) Hemoglobin transports oxygen from the lungs to the other parts of the body.
This animal has oxygen transported from its lungs to other parts of its body.
This animal has hemoglobin in its blood.

Now we get an obviously invalid argument. In order to have a standard deductive explanation of the Hempel variety, the particular explanatory fact cited in the explanans must, in the presence of the general laws in the explanans, be a sufficient condition for the explanandum event. In the hemoglobin case the explanatory fact is a necessary condition of the explanandum event, and that is the typical situation with functional explanations.

Although the suggestion that we admit that necessary conditions can have explanatory import might seem natural at this point, it meets with an immediate difficulty. We would all agree that the fact that a boy was born does not contribute in the least to the explanation of his becoming a juvenile delinquent, but it certainly was a necessary condition of that occurrence. Thus, in general necessary conditions cannot be taken to have any explanatory function, but it is still open to us to see whether...
under certain special conditions necessary conditions can function in
genuine explanations. The approach I am suggesting enables us to do
just that.

Returning to the paresis example, suppose that medical science could
discover in the victims of untreated latent syphilis some characteristic $P$
that is present in each victim who develops paresis and is lacking in each
one who escapes paresis. Then, in the reference class $S$ of untreated
latent syphilitics, we could make a partition in which every member of
one subclass would develop paresis, $B$, whereas no member of the other
subclass would fall victim to it. In such a case, within $S$, $P$ would be a
necessary and sufficient condition for $B$. If, however, medical research
were slightly less successful and found a characteritic $Q$ that is present
in almost every member of $S$ who develops paresis, but is lacking in
almost every one who escapes paresis, then we would still have a
highly relevant partition of the original reference class into two sub-
classes, and this would be the statistical analogue of the discovery of
necessary and sufficient conditions. Thus, if

$$P(S,B) = p_1; \quad P(S,Q,B) = p_2; \quad P(S,\bar{Q},B) = p_3$$

where

$$p_1 \neq p_2 \text{ and } p_1 \neq p_3,$$

then $Q$ is a statistically relevant statistical analogue of a necessary
condition. To concern ourselves with the homogeneity of $S, \bar{Q}$ is to raise the
question of further statistically relevant necessary conditions.

I have argued above that sufficient conditions have explanatory import
only if they are relevant. Insertion of a sodium salt into a Bunsen flame is
a relevant sufficient condition of the flame turning yellow; taking birth
control pills is an irrelevant sufficient condition for nonpregnancy in a
man. Similarly, being born is an irrelevant necessary condition for a boy
becoming delinquent, for it makes no partition whatever in the reference
class of boys, let alone a relevant one. However, as argued above, latent
untreated syphilis is a relevant necessary condition of paresis, and it has
genuine explanatory import.

Let us see whether such considerations are of any help in dealing with
functional explanations. According to classical Freudian theory, dreams
are functionally explained as wish fulfillments. For instance, one night,
after a day of fairly frustrating work on a review of a book containing an
example about dogs who manifest the Absolute by barking loudly, I
dreamed that I was being chased by a pack of barking dogs. Presumably,
my dream had the function of enabling me to fulfill my wish to escape
from the problem of the previous day, and thus to help me preserve my
sleep. The psychoanalytic explanation does not maintain that only one
particular dream would have fulfilled this function; presumably many
others would have done equally well. Thus, the fact that I had that
particular wish did not entail (or even make it very probable) that I
would have that particular dream, but the fact that I had the dream was,
alertly, a sufficient condition for the fulfillment of the wish. The
explanatory fact, the unfulfilled wish, is therefore taken to be a necessary
condition of the explanandum event—the dream—but not a sufficient
condition. According to the psychoanalytic theory, the unfulfilled wish is
a relevant necessary condition, however, for if we make a partition of the
general reference class of nights of sleep in terms of the events of the
preceding day, then the probability of this particular dream is much
greater in the narrower reference class than it was in the original
reference class. By the way, the author of the book I was reviewing is
Stephen Barker.

When we try to give a functional explanation of a particular mecha-
nism, such as the presence of hemoglobin or the occurrence of a dream,
we may feel some dissatisfaction if we are unable eventually to explain
why that particular mechanism occurred rather than another which
would apparently have done the same job equally well. Such a feeling is,
I would think, based upon the same considerations that lead us to feel
dissatisfied with the explanation of paresis. We know some necessary
conditions, in all of these cases, but we believe that it is possible in
principle to find sufficient conditions as well, and we will not regard the
explanations as complete until we find them. How well this feeling is
justified depends chiefly upon the field of science in which the explana-
tion is sought, as the example of radioactive decay of a $U^{238}$ atom shows.

The fact that the multiple homogeneity rule allows for the explanatory
value of relevant necessary conditions and their statistical analogues
does not, by itself, provide a complete account of functional explanation,
for much remains to be said about the kinds of necessary conditions that
enter into functional explanations and their differences from necessary
conditions that are not in any sense functional. Our approach has,
however, removed what has seemed to many authors the chief stumbling
block in the path to a satisfactory theory of functional explanation.

10. Explanation Without Increase of Weight

It is tempting to suppose, as I have been doing so far and as all the
examples have suggested, that explanation of an explanandum event
somehow confers upon it a posterior weight that is greater than its prior weight. When I first enunciated this principle in section 3, however, I indicated that, though heuristically beneficial, it should not be considered fully accurate. Although most explanations may conform to this general principle, I think there may be some that do not. It is now time to consider some apparent exceptions and to see whether they are genuine exceptions.

Suppose, for instance, that a game of heads and tails is being played with two crooked pennies, and that these pennies are brought in and out of play in some irregular manner. Let one penny be biased for heads to the extent that 90 percent of the tosses with it yield heads; let the other be similarly biased for tails. Furthermore, let the two pennies be used with equal frequency in this game, so that the overall probability of heads is one-half. (Perhaps a third penny, which is fair, is tossed to decide which of the two biased pennies is to be used for any given play.) Suppose a play of this game results in a head; the prior weight of this event is one-half. The general reference class of plays can, however, be partitioned in a statistically relevant way into two homogeneous reference classes. If the toss was made with the penny biased for heads, the result is explained by that fact, and the weight of the explanandum event is raised from 0.5 to 0.9.

Suppose, however, that the toss were made with the penny biased for tails; the explanandum event is now referred to the other subclass of the original reference class, and its weight is decreased from 0.5 to 0.1. Do we want to say in this case that the event is thereby explained? Many people would want to deny it, for such cases conflict with their intuitions (which, in many cases, have been significantly conditioned by Hempel's persuasive treatment) about what an explanation ought to be. I am inclined, on the contrary, to claim that this is genuine explanation. There are, after all, improbable occurrences—such events are not explained by making them probable. Is it not a peculiar prejudice to maintain that only those events which are highly probable are capable of being explained—that improbable events are in principle inexplicable? Any event, regardless of its probability, is amenable to explanation, I believe; in the case of improbable events, the correct explanation is that they are highly improbable occurrences which happen, nevertheless, with a certain definite frequency. If the reference class is actually homogeneous, there are no other circumstances with respect to which they are probable. No further explanation can be required or can be given.

There are various reasons for which my view might be rejected. In the first place, I am inclined to think that the deterministic prejudice may often be operative—namely, that x is B is not explained until x is incorporated within a reference class all of whose members are B. This is the feeling that seemed compelling in connection with the paresis example. In the discussion of that example, I indicated my reasons for suggesting that our deterministic hopes may simply be impossible to satisfy. In an attempt to undercut the hope generated in medical science that determinism would eventually triumph, I introduced the parallel example of the explanation of a radioactive decay in an alloy of lead and uranium, in which case there is strong reason to believe that the reference class of uranium-238 atoms is strictly homogeneous with respect to disintegration.

In order to avoid being victimized by the same deterministic hope in the present context, we could replace the coin-tossing example at the beginning of this section with another example from atomic physics. For instance, we could consider a mixture of uranium-235 atoms, whose half-life is 4.5 x 10^7 years, and polonium-214 atoms, whose half-life is 1.6 x 10^8 seconds. The probability of disintegration of an unspecified atom in the mixture is between that for atoms of U-235 and Po-214. Suppose that within some small specified time interval a decay occurs. There is a high probability of a polonium atom disintegrating within that interval, but a very low probability for a uranium atom. Nevertheless, a given disintegration may be of a uranium atom, so the transition from the reference class of a mixture of atoms of the two types to a reference class of atoms of U-235 may result in a considerable lowering of the weight. Nevertheless, the latter reference class may be unqualifiedly homogeneous. When we ask why that particular atom disintegrated, the answer is that it was a U-235 atom, and there is a small probability that such an atom will disintegrate in a short time interval.

If, in the light of modern developments in physics and philosophy, determinism no longer seems tenable as an a priori principle, we may try to salvage what we can by demanding that an explanation that does not necessitate its explanandum must at least make it highly probable. This is what Hempel's account requires. I have argued above, in the light of various examples, that even this demand is excessive and that we must accept explanations in which the explanandum event ends up with a low posterior weight. "Well, then," someone might say, "if the explanation does not show us that the event was to be expected, at least it ought to show us that the event was to be expected somewhat more than it otherwise would have been." But this attitude seems to derive in an
unfortunate way from regarding explanations as arguments. At this juncture it is crucial to point out that the emphasis in the present account of explanation is upon achieving a relevant partition of an inhomogeneous reference class into homogeneous subclasses. On this conception an explanation is not an argument that is intended to produce conviction; instead, it is an attempt to assemble the factors that are relevant to the occurrence of an event. There is no more reason to suppose that such a process will increase the weight we attach to such an occurrence than there is to suppose that it will lower it. Whether the posterior weight is higher than or lower than the prior weight is really beside the point. I shall have more to say later about the function of explanations.

Before leaving this topic, I must consider one more tempting principle. It may seem evident from the examples thus far considered that an explanation must result in a change—an increase or a decrease—in the transition from the prior weight to the posterior weight. Even this need not occur. Suppose, for instance, that we change the coin-tossing game mentioned at the outset of this section by introducing a fair penny into the play. Now there are three pennies brought into play randomly: one with a probability of 0.9 for heads, one with a probability of 0.5 for heads, and one with a probability of 0.1 for heads. Overall, the probability of heads in the game is still one-half. Now, if we attempt to explain a given instance of a head coming up, we may partition the original reference class into three homogeneous subclasses, but in one of these three the probability of heads is precisely the same as it is in the entire original class. Suppose our particular head happens to belong to that subclass. Then its prior weight is exactly the same as its posterior weight, but I would claim that explanation has occurred simply by virtue of the relevant partition of the original nonhomogeneous reference class. This makes sense if one does not insist upon regarding explanations as arguments.

11. Some Paradigms of Explanation

Before attempting a general quasi-formal characterization of the explanation of particular events, I should like to follow a common practice and introduce some examples that seem to me to deserve the status of paradigms of explanation. They will, I believe, differ in many ways from the paradigms that are usually offered; they come from a set of investigations, conducted mainly by Grünbaum and Reichenbach, concerning the temporal asymmetry (or anisotropy) of the physical world. Because of the close connections between causality and time on the one hand, and

between causality and explanation on the other, these investigations have done a great deal to elucidate the problem of explanation.

Given a thermally isolated system, there is a small but nonvanishing probability that it will be found in a low entropy state. A permanently closed system will from time to time, but very infrequently, spontaneously move from a state of high entropy to one of low entropy, although by far the vast majority of its history is spent in states of high entropy. Let us take this small probability as the prior probability that a closed system will be in a low entropy state. Suppose, now, that we examine such a system and find that it is in a low entropy state. What is the explanation? Investigation reveals that the system, though closed at the time in question, had recently interacted with its environment. The low entropy state is explained by this interaction. The low entropy of the smaller system is purchased by an expenditure of energy in the larger environmental system that contains it and by a consequent increase in the entropy of the environment.

For example, there is a small but nonvanishing probability that an ice cube will form spontaneously in a thermos of cool water. Such an occurrence would be the result of a chance clustering of relatively nonenergetic molecules in one particular place in the container. There is a much greater chance that an ice cube will be present in a thermos of cool water to which an ice cube has just been added. Even if we had no independent knowledge about an ice cube having been added, we would confidently infer it from the mere presence of the ice cube. The low entropy state is explained by a previous interaction with the outside world, including a refrigerator that manufactures the ice cubes.

Suppose, for a further example, that we found a container with two compartments connected by an opening, one compartment containing only oxygen and the other only nitrogen. In this case, we could confidently infer that the two parts of the container had been separately filled with oxygen and nitrogen and that the connecting window had recently been opened—so recently that no diffusion had yet taken place. We would not conclude that the random motions of the molecules had chanced to separate the two gases into the two compartments, even though that event has a small but nonvanishing probability. This otherwise improbable state of affairs is likewise explained in terms of a recent interaction with the environment.

In examples of these kinds, the antecedent reference class of states of closed systems is recognized as inhomogeneous, and it is partitioned into those states which follow closely upon an interaction with the outside
world and those that do not. Within the former homogeneous subclass of the original reference class, the probability of low entropy states is much higher than the probability of such states referred to the unpartitioned inhomogeneous reference class. In such cases, we, therefore, have explanations of the low entropy states.

The foregoing examples involve what Reichenbach calls “branch systems.” In his extended and systematic treatment of the temporal asymmetry of the physical world, he accepts it as a fundamental fact that our region of the universe has an abundant supply of these branch systems. During their existence as separate systems they may be considered closed, but they have not always been isolated, for each has a definite beginning to its history as a closed system. If we take one such system and suppose it to exist from then on as a closed system, we can consider its successive entropy states as a probability sequence, and the probability of low entropy in such a sequence is very low. This probability is a one-system probability referred to a time ensemble; as before, let us take it as a prior probability. If we examine many such systems, we find that a large percentage of them are in low entropy states shortly after their inceptions as closed systems. There is, therefore, a much higher probability that a very young branch system is in a low entropy state. This is the posterior probability of a low entropy state for a system that has recently interacted with its environment; it is a many-system probability referred to a space ensemble. The time ensemble turns out to be an inhomogeneous reference class; the space ensemble yields homogeneous reference classes. Applying this consideration to our ice cube example, we see that the presence of the ice cube in the thermos of cool water is explained by replacing the prior weight it would have received from the reference class of states of this system as an indefinitely isolated container of water with the weight it receives from the reference class of postinteraction states of many such containers of water when the interaction involves the placing of an ice cube within it.

Even if we recognize that the branch systems have not always existed as closed systems, it is obviously unrealistic to suppose that they will remain forever isolated. Instead, each branch system exists for a finite time, and at each temporal end it merges with its physical environment. Each system exhibits low entropy at one temporal end and high entropy at the other. The basic fact about the branch systems is that the vast majority of them exhibit low entropy states at the same temporal end. We can state this fact without making any commitment whatever as to whether that end is the earlier or the later end. However, the fact that the systems are alike in having low entropy at the same end can be used to provide coordinating definitions of such temporal relations as earlier and later. We arrive at the usual temporal nomenclature if we say that the temporal end at which the vast majority of branch systems have low entropy is the earlier end and the end at which most have high entropy is the later end. It follows, of course, that the interaction with the environment at the low entropy end precedes the low entropy state.

Reichenbach takes this relationship between the initial interaction of the branch system with its environment and the early low entropy state to be the most fundamental sort of producing. The interaction produces the low entropy state, and the low entropy state is the product (and sometimes the record) of the interaction. Producing is, of course, a thoroughly causal concept; thus, the relation of interaction to orderly low entropy state becomes the paradigm of causation on Reichenbach’s view, and the explanation of the ice cube in the thermos of water is an example of the most fundamental kind of causal explanation. The fact that this particular interaction involves a human act is utterly immaterial, for many other examples exist in which human agency is entirely absent.

Reichenbach’s account, if correct, shows why causal explanation is temporally asymmetrical. Orderly states are explained in terms of previous interactions; the interactions that are associated with low entropy states do not follow them. It appears that the temporal asymmetry of causal explanation is preserved even when the causal concepts are extended to refer to reversible mechanical processes. It is for this reason, I imagine, that we are willing to accept an explanation of an eclipse in terms of laws of motion and antecedent initial conditions, but we feel queasy, to say the least, about claiming that the same eclipse can be explained in terms of the laws of motion and subsequent initial conditions, even though we may infer the occurrence of the eclipse equally well from either set of initial conditions. I shall return to this question of temporal asymmetry of explanation in the next section.

The foregoing examples are fundamentally microstatistical. Reichenbach has urged that macrostatistical occurrences can be treated analogously. For instance, a highly ordered arrangement in a deck of cards can be explained by the fact that it is newly opened and was arranged at the factory. Decks of cards occasionally get shuffled into highly ordered arrangements—just as closed thermodynamic systems occasionally develop low entropy states—but much more frequently the order arises from a deliberate arrangement. Reichenbach likewise argues that common causes macrostatistically explain improbable “coincidences.” For
example, suppose all the lights in a certain block go off simultaneously. There is a small probability that in every house the people decided to go to bed at precisely the same instant. There is a small probability that all the bulbs burned out simultaneously. There is a much greater probability of trouble at the power company or in the transmission lines. When investigation reveals that a line was downed by heavy winds, we have an explanation of a coincidence with an extremely low prior weight. A higher posterior weight results when the coincidence is assigned to a homogeneous reference class. The same considerations apply, but with much more force, when it is the entire eastern seaboard of the United States that suffers a blackout.

Reichenbach believed that it is possible to establish temporal asymmetry on the basis of macrostatistical relations in a manner quite analogous to the way in which he attempted to base it upon microstatistical facts. His point of departure for this argument is the fact that certain improbable coincidences occur more often than would be expected on the basis of chance alone. Take, as an example, all members of a theatrical company falling victim to a gastrointestinal illness on the same evening. For each individual there is a certain probability (which may, of course, vary from one person to another) that he will become ill on a given day. If the illnesses of the members of the company are independent of one another, the probability of all members of the company becoming ill on the same day is simply the product of all the separate probabilities; for the general multiplication rule says (for two events) that

\[ P(A, B, C) = P(A, B) \times P(A, B, C), \]

but if

\[ P(A, C) = P(A, B, C), \]

we have the special multiplication rule,

\[ P(A, B, C) = P(A, B) \times P(A, C). \]

When relation (3) holds, we say that the events B and C are independent of each other; this is equivalent to relation (2), which is our definition of statistical irrelevance.

We find, as a matter of fact, that whole companies fall ill more frequently than would be indicated if the separate illnesses were all statistically independent of one another. Under these circumstances, Reichenbach maintains, we need a causal explanation for what appears to be an improbable occurrence. In other words, we need a causal explanation for the statistical relevance of the illnesses of the members of the company to one another. The causal explanation may lie in the food that all of the members of the company ate at lunch. Thus, we do not assert a direct causal relation between the illness of the leading man and that of the leading lady; it was not the fact that the leading man got sick that caused the illness of the leading lady (or the actors with minor roles, the stand-ins, the prop men, etc.). There is a causal relation, but it is via a common cause—the spoiled food. The common meal screens off the statistical relevance of the illness of the leading man to that of the leading lady. The case is parallel to that of the barometer and the storm, and the screening off works in the same way.

Reichenbach maintains that coincidences of this sort can be explained by common causes. The illness of the whole company is explained by the common meal; the blackout of an entire region is explained by trouble at the power source or in the distribution lines. If we take account of the probability of spoiled food's being served to an entire group, the frequency with which the members all become victims of a gastrointestinal illness on the same day is not excessive. If we take into account the frequency with which power lines go down or of trouble at the distribution station, etc., the frequency with which whole areas simultaneously go dark is not excessive.

Such occurrences are not explained in terms of common effects. We do not seriously claim that all members of the company became ill because events were conspiring (with or without conscious intent of some agency, natural or supernatural) to bring about a cancellation of the performance; we do not believe that the future event explains the present illness of the company. Similarly, we do not explain the blackout of the eastern seaboard in terms of events conspiring to make executives impregnate their secretaries. The situation is this: in the absence of a common cause, such as spoiled food, the cancellation of the play does not make the coincidental illness of the company any more probable than it would have been as a product of probabilities of independent events.

By contrast, the presence of a common cause such as the common meal does make the simultaneous illness more probable than it would have been as a product of independent events, and that is true whether or not the common effect occurs. The efficacy of the common cause is not affected by the question of whether the play can go on, with a sick cast or with substitutes. The probability of all the lights going out simultaneously, in the absence of a common cause, is unaffected by the question of how the men and women, finding themselves together in the dark for the
entire night without spouses, behave themselves. The net result is that
coincidences of the foregoing type are to be explained in terms of causal
antecedents, not subsequent events. The way to achieve a relevant
subdivision of the general reference class is in terms of antecedent
conditions, not later ones. This seems to be a pervasive fact about our
macroworld, just as it seems to characterize the microworld.

I believe that an important feature of the characterization of explanation
I am offering is that it is hospitable to paradigms of the two kinds I
have discussed in this section, the microstatistical examples and the
macrostatistical examples. In all these cases, there is a prior probability
of an event which we recognize as furnishing an inappropriate weight
for the event in question. We see that the reference class for the prior
probability is inhomogeneous and that we can make a relevant partition.
The posterior probability which arises from the new homogeneous refer-
ence class is a suitable weight to attach to the single event. It is the
introduction of statistically relevant factors for partitioning the reference
class that constitutes the heart of these explanations. If these examples
are, indeed, paradigms of the most fundamental types of explanation, it
is a virtue of the present account that it handles them easily and
naturally.

12. The Temporal Asymmetry of Explanation

It is also a virtue of the present account, I believe, that it seems to
accommodate the rather puzzling temporal asymmetry of explanation
already noted above—the fact that we seem to insist upon explaining
events in terms of earlier rather than later initial conditions. Having
noted that both the microstatistical approach and the macrostatistical
approach yield a fundamental temporal asymmetry, let us now use that
fact to deal with a familiar example. I shall attempt to show how
Reichenbach's principle of the common cause, introduced in connection
with the macrostatistical examples discussed above, helps us to establish
the temporal asymmetry of explanation.

Consider Silvain Bromberger's flagpole example, which goes as fol-
loows. On a sunny day a flagpole of a particular height casts a shadow of
some particular length depending upon the elevation of the sun in the
sky. We all agree that the position of the sun and the height of the
flagpole explain the length of the shadow. Given the length of the
shadow, however, and the position of the sun in the sky, we can equally
infer the height of the flagpole, but we rebel at the notion that the length
of the shadow explains the height of the flagpole. It seems to me that the

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temporal relations are crucial in this example. Although the sun, flagpole,
and shadow are perhaps commonsensically regarded as simultaneous, a
more sophisticated analysis shows that physical processes going on in
time are involved. Photons are emitted by the sun, they travel to the
vicinity of the flagpole where some are absorbed and some are not, and
those which are not go on to illuminate the ground. A region of the
ground is not illuminated, however, because the photons traveling
toward that region were absorbed by the flagpole. Clearly the interac-
tion between the photons and the flagpole temporally precedes the interac-
tion between the neighboring photons and the ground. The reason that
the explanation of the length of the shadow in terms of the height of the
flagpole is acceptable, whereas the "explanation" of the height of the
flagpole in terms of the length of the shadow is not acceptable, seems to
me to hinge directly upon the fact that there are causal processes with
earlier and later temporal stages. It takes only a very moderate extension
of Reichenbach's terminology to conclude that the flagpole produces the
shadow in a sense in which the shadow certainly does not produce the
flagpole.

If we give up the notion that explanations are arguments, there is no
need to be embarrassed by the fact that we can oftentimes infer earlier
events from later ones by means of nomological relations, as in the cases
of the eclipse and the flagpole. I have been arguing that relevance
considerations are preeminent for explanations; let us see whether this
approach provides a satisfactory account of the asymmetry of explana-
tion in the flagpole case. The apparent source of difficulty is that, under
the general conditions of the example, the flagpole is relevant to the
shadow and the shadow is relevant to the flagpole. It does not follow,
however, that the shadow explains the flagpole as well as the flagpole
explains the shadow.

In order to analyze the relevance relations more carefully, I shall
reevaluate a simplified version of the example of the illness in the
theatrical company and compare it with a simplified version of the
flagpole example. In both cases the screening-off relation will be em-
ployed in an attempt to establish the temporal asymmetry of the explana-
tion. The general strategy will be to show that a common cause screens
off a common effect and, consequently, by using the screening-off rule,
that the explanation must be given in terms of the common cause and not
in terms of the common effect. In order to carry out this plan, I shall
regard the flagpole as an orderly arrangement of parts that requires an
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explanation of the same sort as does the coincidental illnesses of the actors.

Consider, then, the simple case of a theatrical company consisting of only two people, the leading lady and the leading man. Let A be our general reference class of days, let M be the illness of the leading man, let L be the illness of the leading lady, let F be a meal of spoiled food that both eat, and let C be the cancellation of the performance. Our previous discussion of the example has shown that the simultaneous illness occurs more frequently than it would if the two were independent, that is,

\[ P(A,L,M) > P(A,L) \times P(A,M), \]

and that the common meal is highly relevant to the concurrent illnesses, that is,


It is not true, without further qualification, that the eating of spoiled food F screens off the cancellation of the performance C from the joint illness L.M. Although the eating of spoiled food makes it probable that the two actors will be ill, the fact that the performance has been cancelled supplies further evidence of the illness and makes the joint illness more probable. At this point experiment must be admitted. It is clearly possible to arrange things so that the play goes on, illness or no illness, by providing substitutes who never eat with the regular cast. Likewise, it is a simple matter to see to it that the performance is cancelled, whether or not anyone is ill. Under these experimental conditions we can see that the probability of the joint illness, given the common meal with spoiled food, is the same whether or not the play is performed, that is,


From this it follows that


If we approach the common meal of spoiled food in the same experimental fashion, we can easily establish that it is not screened off by the cancellation of the performance. We can arrange for the two leading actors to have no meals supplied from the same source and ascertain the probability \( P(A,F,C,L,M) \). This can be compared with the probability \( P(A,F,C,L,M) \) which arises under the usual conditions of the two actors eating together in the same restaurants. Since they are not equal, the common spoiled food is statistically relevant to the joint illness, even in the presence of the cancellation of the performance, that is,

\[ P(A,F,C,L,M) \neq P(A,F,L,M). \]

Thus, although the common cause F is relevant to the coincidence L.M and the coincidence L.M is relevant to C (from which it follows that C is relevant to L.M), it turns out that the common cause F screens off the common effect C from the coincidence to be explained. By the screening-off rule, F must be used and C must not be used to partition the reference class A. These considerations express somewhat formally the fact that tampering with the frequency of F without changing the frequency of C will affect the frequency of L.M, but tampering with the frequency of C without changing the frequency of F will have no effect on the frequency of L.M. This seems to capture the idea that we can influence events by influencing their causal antecedents, but not by influencing their causal consequents.

Let us apply the same analysis to the flagpole example. Again, let A be the general reference class, and let us for simplicity suppose that the flagpole is composed of two parts, a top and a bottom. The flagpole is in place when the two parts are in place; let T be the proper positioning of the top, and B the proper positioning of the bottom. Let M represent the flagpole makers’ bringing the pieces together in the appropriate positions. Let S be the occurrence of the full shadow of the flagpole. Now, since the flagpole’s existence consists in the two pieces’ being put together in place and since that hardly happens except when the flagpole makers bring them together and assemble them, it is clear that

\[ P(A,T,B) > P(A,T) \times P(A,B). \]

Hence, the existence of the flagpole is something to be explained. Since we know that


M effects a relevant partition in the general reference class A.

The question of whether the flagpole gets put together when the flagpole makers go about putting it together is unaffected by the existence or nonexistence of a shadow—for example, the ground might be illuminated by other sources of light besides the sun, or mirrors might deflect some of the sun’s rays, with the result that there is no shadow—but the flagpole is there just the same; hence,
This, therefore, conclude again that the common cause screens off the common effect and
that, by virtue of the screening-off rule, the causal antecedent $M$—and
not the causal consequent $S$—must be invoked to effect a relevant
partition in the reference class $A$. In this highly schematic way, I hope I
have shown that the approach to explanation via statistical relevance has
allowed us to establish the temporal asymmetry of explanation. Once
more, the intuitive idea is that manipulating the pieces of the flagpole,
without otherwise tampering with the shadow, affects the shadow; con­
trariwise, tampering with the shadow, without otherwise manipulating
the pieces of the flagpole, has no effect upon the flagpole. The analysis of
the flagpole example may, of course, require the same sort of appeal to
experiment as we invoked in the preceding example.

I hope it is obvious from all that has been said that none of us is
claiming that the temporal asymmetries herein discussed are anything
beyond very general facts about the world. Reichenbach regards it as a
matter of fact, not of logical necessity, that common causes have their
particular statistical relations to the "coincidences" they explain. In a
different universe it might be quite otherwise. Neither Reichenbach nor
Griinbaum regards the entropic behavior of branch systems as logical
necessities, so they are both prepared to deny any logical necessity to the
fact that low entropy states are explained by previous interactions.
Again, in a different universe, or in another part of this one, it might be
different. Similarly, when I argue that explanations from earlier initial
conditions are acceptable, whereas explanations from later initial condi­
tions are inadmissible, I take this to be an expression of certain rather
general facts about the world. I certainly do not take this characteristic
of explanation to be a matter of logical necessity.

The general fact about the world that seems to be involved is that
causal processes very frequently exhibit the following sort of structure:
a process leading up to a given event $E$ consists of a series of events
earlier than $E$, but such that later ones screen off earlier ones. In other
words, a given antecedent event $A_1$ will be relevant to $E$, but it will be
screened off by a later antecedent $A_2$ that intervenes between $A_1$ and $E$.
This situation obtains until we get to $E$, and then every subsequent event
is screened off by some causal antecedent or other. Thus, in some deeply
significant sense, the causal consequents of an event are made irrelevant
to its occurrence in a way in which the causal antecedents are not. If, in
Velikovsky-like cataclysm, a giant comet should disrupt the solar system,
it would have enormous bearing upon subsequent eclipses but none what­
ever upon previous ones. The working out of the details of the eclipse
example, along the lines indicated by the analysis of the flagpole example,
is left as an exercise for the reader.

To say that causal antecedents can affect an event in ways in which
causal consequents cannot is, of course, an utter banality. The difficulty
is that causal consequents are not statistically irrelevant to an event, so it
is rather problematic to state with any precision the kind of asymmetry
of relevance we are trying to capture. Reichenbach, himself, is quite
careful to point out that the notion of influencing future events is
infected with a fundamental temporal ambiguity. I am not at all confident
that the use of the screening-off relation is sufficient for a complete account of the relevance relations we seek, but it does have a
promising kind of asymmetry, and so I am led to hope that it will
contribute at least part of the answer.

13. The Nature of Statistical Explanation

Let me now, at long last, offer a general characterization of explana­
tions of particular events. As I have suggested earlier, we may think of
an explanation as an answer to a question of the form, "Why does this $x$
which is a member of $A$ have the property $B$?" The answer to such a
question consists of a partition of the reference class $A$ into a number of
subclasses, all of which are homogeneous with respect to $B$, along with
the probabilities of $B$ within each of these subclasses. In addition, we
must say which of the members of the partition contains our particular $x$.
More formally, an explanation of the fact that $x$, a member of $A$, is a mem­
ber of $B$ would go as follows:

\[
P(A.C_1,B) = p_1
\]
\[
P(A.C_2,B) = p_2
\]
\[
\vdots
\]
\[
P(A.C_n,B) = p_n
\]

where
The problem is to select an appropriate value for \( r \). No value is satisfactory, for no matter how large \( r \) is, provided it is less than one, we can construct a fair lottery with a sufficient number of tickets to be able to say for each ticket that will not win, because the probability of its not winning is greater than \( r \). From this we can conclude that no ticket will win, which contradicts the stipulation that this is a fair lottery— no lottery can be considered fair if there is no winning ticket.

It was an exceedingly profound insight on Carnap’s part to realize that inductive logic can, to a large extent anyway, dispense entirely with rules of acceptance and inductive inferences in the ordinary sense. Instead, inductive logic attaches numbers to hypotheses, and these numbers are used to make practical decisions. In some circumstances such numbers, the degrees of confirmation, may serve as fair betting quotients to determine the odds for a fair bet on a given hypothesis. There is no rule that tells one when to accept an hypothesis or when to reject it; instead, there is a rule of practical behavior that prescribes that we so act as to maximize our expectation of utility. 

Hence, inductive logic is simply not concerned with inductive arguments (regarded as entities composed of premises and conclusions). Now, I do not completely agree with Carnap on the issue of acceptance rules in inductive logic; I believe that inductive logic does require some inductive inferences. But when it comes to probabilities (weights) of single events, I believe that he is entirely correct. In my view, we must establish by inductive inference probability statements, which I regard as statements about limiting frequencies. But, when we come to apply this probability knowledge to single events, we procure a weight which functions just as Carnap has indicated—as a fair betting quotient or as a value to be used in computing an expectation of utility. Consequently, I maintain, in the context of statistical explanation of individual events, we do not need to try to establish the explanandum as the conclusion of an inductive argument; instead, we need to establish the weights that would appropriately attach to such explanandum events for purposes of betting and other practical behavior. That is precisely what the partition of the reference class into homogeneous subclasses achieves: it establishes the correct weight to assign to any member of \( A \) with respect to its being \( B \). First, one determines to which compartment \( C_i \) it belongs, and then one adopts the value \( p_i \) as the weight. Since we adopted the multiple homogeneity rule, we can genuinely handle any member of \( A \), not just those which happen to fall into one subclass of the original reference class.

One might ask on what grounds we can claim to have characterized explanation. The answer is this. When an explanation (as herein explicated) has been provided, we know exactly how to regard any \( A \) with respect to the property \( B \). We know which ones to bet on, which to bet against, and at what odds. We know precisely what degree of expectation is rational. We know how to face uncertainty about an \( A \)'s being a \( B \) in the most reasonable, practical, and efficient way. We know every factor that is relevant to an \( A \) having property \( B \). We know exactly the weight that should have been attached to the prediction that this \( A \) will be a \( B \). We know all of the regularities (universal or statistical) that are relevant to our original question. What more could one ask of an explanation?

There are several general remarks that should be added to the foregoing theory of explanation:

a. It is evident that explanations as herein characterized are nomological. For the frequency interpretation probability statements are statistical generalizations, and every explanation must contain at least one such generalization. Since an explanation essentially consists of a set of statistical generalizations, I shall call these explanations "statistical" without
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qualifications, meaning thereby to distinguish them from what Hempel has recently called "inductive-statistical." His inductive-statistical explanations contain statistical generalizations, but they are inductive inferences as well.

b. From the standpoint of the present theory, deductive-nomological explanations are just a special case of statistical explanation. If one takes the frequency theory of probability as literally dealing with infinite classes of events, there is a difference between the universal generalization, "All A are B," and the statistical generalization, "P(A,B) = 1," for the former admits no As that are not Bs, whereas the latter admits of infinitely many As that are not Bs. For this reason, if the universal generalization holds, the reference class A is homogeneous with respect to B, whereas the statistical generalization may be true even if A is not homogeneous. Once this important difference is noted, it does not seem necessary to offer a special account of deductive-nomological explanations.

c. The problem of symmetry of explanation and prediction, which is one of the most hotly debated issues in discussions of explanation, is easily answered in the present theory. To explain an event is to provide the best possible grounds we could have had for making predictions concerning it. An explanation does not show that the event was to be expected; it shows what sorts of expectations would have been reasonable and under what circumstances it was to be expected. To explain an event is to show to what degree it was to be expected, and this degree may be translated into practical predictive behavior such as wagering on it. In some cases the explanation will show that the explanandum event was to be expected, but that does not destroy the symmetry of explanation and prediction. The symmetry consists in the fact that the explanatory facts constitute the fullest possible basis for making a prediction of whether or not the event would occur. To explain an event is not to predict it ex post facto, but a complete explanation does provide complete grounds for rational prediction concerning that event. Thus, the present account of explanation does sustain a thoroughlygoing symmetry thesis, and this symmetry is not refuted by explanations having low weights.

d. In characterizing statistical explanation, I have required that the partition of the reference class yield subclasses that are, in fact, homogeneous. I have not settled for practical or epistemic homogeneity. The question of whether actual homogeneity or epistemic homogeneity is demanded is, for my view, analogous to the question of whether the

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premises of the explanation must be true or highly confirmed for Hempel's view. I have always felt that truth was the appropriate requirement, for I believe Carnap has shown that the concept of truth is harmless enough. However, for those who feel too uncomfortable with the stricter requirement, it would be possible to characterize statistical explanation in terms of epistemic homogeneity instead of actual homogeneity. No fundamental problem about the nature of explanation seems to be involved.

e. This paper has been concerned with the explanation of single events, but from the standpoint of probability theory, there is no significant distinction between a single event and any finite set of events. Thus, the kind of explanation appropriate to a single result of heads on a single toss of a coin would, in principle, be just like the kind of explanation that would be appropriate to a sequence of ten heads on ten consecutive tosses of a coin or to ten heads on ten different coins tossed simultaneously.

f. With Hempel, I believe that generalizations, both universal and statistical, are capable of being explained. Explanations invoke generalizations as parts of the explanans, but these generalizations themselves may need explanation. This does not mean that the explanation of the particular event that employed the generalization is incomplete; it only means that an additional explanation is possible and may be desirable. In some cases it may be possible to explain a statistical generalization by subsuming it under a higher level generalization; a probability may become an instance for a higher level probability. For example, Reichenbach offered an explanation for equiprobability in games of chance, by constructing, in effect, a sequence of probability sequences. Each of the first level sequences is a single case with respect to the second level sequence. To explain generalizations in this manner is simply to repeat, at a higher level, the pattern of explanation we have been discussing. Whether this is or is not the only method of explaining generalizations is, of course, an entirely different question.

g. In the present account of statistical explanation, Hempel's problem of the "nonconjunctiveness of statistical systematization" simply vanishes. This problem arises because in general, according to the multiplication theorem for probabilities, the probability of a conjunction is smaller than that of either conjunct taken alone. Thus, if we have chosen a value r, such that explanations are acceptable only if they confer upon the explanandum an inductive probability of at least r, it is quite possible that each of the two explananda will satisfy that condition, whereas their
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conjunction fails to do so. Since the characterization of explanation I am offering makes no demands whatever for high probabilities (weights), it has no problem of nonconjunctiveness.

14. Conclusion

Although I am hopeful that the foregoing analysis of statistical explanation of single events solely in terms of statistical relevance relations is of some help in understanding the nature of scientific explanation, I should like to cite, quite explicitly, several respects in which it seems to be incomplete.

First, and most obviously, whatever the merits of the present account, no reason has been offered for supposing the type of explanation under consideration to be the only legitimate kind of scientific explanation. If we make the usual distinction between empirical laws and scientific theories, we could say that the kind of explanation I have discussed is explanation by means of empirical laws. For all that has been said in this paper, theoretical explanation—explanation that makes use of scientific theories in the fullest sense of the term—may have a logical structure entirely different from that of statistical explanation. Although theoretical explanation is almost certainly the most important kind of scientific explanation, it does, nevertheless, seem useful to have a clear account of explanation by means of empirical laws, if only as a point of departure for a treatment of theoretical explanation.

Second, in remarking above that statistical explanation is nomological, I was tacitly admitting that the statistical or universal generalizations invoked in explanations should be lawlike. I have made no attempt to analyze lawlikeness, but it seems likely that an adequate analysis will involve a solution to Nelson Goodman’s “grue-bleen” problem.

Third, my account of statistical explanation obviously depends heavily upon the concept of statistical relevance and upon the screening-off relation, which is defined in terms of statistical relevance. In the course of the discussion, I have attempted to show how these tools enable us to capture much of the involvement of explanation with causality, but I have not attempted to provide an analysis of causation in terms of these statistical concepts alone. Reichenbach has attempted such an analysis,

but whether his—or any other—can succeed is a difficult question. I should be inclined to harbor serious misgivings about the adequacy of my view of statistical explanation if the statistical analysis of causation cannot be carried through successfully, for the relation between causation and explanation seems extremely intimate.

Finally, although I have presented my arguments in terms of the limiting frequency conception of probability, I do believe that the fundamental correctness of the treatment of statistical explanation hinges upon the acceptability of that interpretation of probability. Proponents of other theories of probability, especially the personalist and the propensity interpretations, should be able to adapt this treatment of explanation to their views of probability with a minimum of effort. That, too, is left as an exercise for the reader.

NOTES


3. See also Hempel, Aspects of Scientific Explanation, pp. 357-68, where he offers a “general condition of adequacy for any rationally acceptable explanation of a particular event,” namely, that “any rationally acceptable answer to the question ‘why did event X occur?’ must offer information which shows that X was to be expected—if not definitely, as in the case of D-N explanation, then at least with reasonable probability.”

Inductive explanations have variously been known as “statistical,” “probabilistic,” and “inductive-statistical.” Deductive explanations have often been called “deductive-nomological.” For the present I shall simply use the terms “inductive” and “deductive” to emphasize the crucial fact that the former embody inductive logical relations, whereas the latter embody deductive logical relations. Both types are nomological, for both require lawlike generalizations among their premises. Later on, I shall use the term “statistical explanation” to refer to the sort of explanation I am trying to characterize, for it is statistic in a straightforward sense, and it is noninductive in an extremely important sense.


7. I called attention to this danger in “The Status of Prior Probabilities in Statistical Explanation,” Philosophy of Science, XXXII, no. 2 (April, 1965), p. 137. Several fundamental disanalogies could be cited. First, the relation of deductive entailment is transitive, whereas the relation of inductive support is not; see my “Consistency, Transitivity, and Inductive Support,” Ratio, VII, no. 2 (Dec. 1965), pp. 164-65. Second, on Carnap’s theory of degree of confirmation, which is very close to the notion of inductive probability that Hempel uses in characterizing statistical explanation, there is no such thing as inductive inference in the sense of allowing the detachment of inductive conclusions in a manner analogous
to that in which deductive logic allows the detachment of conclusions of deductive inferences. See my contribution "Who Needs Inductive Acceptance Rules?" to the discussion of Henry E. Kyburg's "The Rule of Detachment in Inductive Logic," in The Problem of Inductive Logic, ed. Imre Lakatos (Amsterdam: North Holland Publishing Co., 1968), pp. 159-164, for an assessment of the bearing of this disanalogy specifically upon the problem of scientific explanation. Third, if q follows from p by a deductively valid argument, then q follows validly from p & r, regardless of what statement r is. This is the reason that the requirement of total evidence is automatically satisfied for deductive-nomological explanations. By contrast, even if p provides strong inductive support for q, q may not be inductively supported at all by p & r. Informally, a valid deductive argument remains valid no matter what premises are added (as long as none is taken away), but addition of premises to a strong inductive argument can destroy all of its strength. It is for this reason that the requirement of total evidence is not vacuous for statistical explanations.

8. See "Deductive-Nomological vs. Statistical Explanation."

9. See, for example, Hempel, "Explanation in Science and in History."

10. In the present context nothing important hinges upon the particular characterization of the parts of arguments. I shall refer to them indifferently as statements or propositions. Propositions may be regarded as classes of statements; so long as they are not regarded as facts of the world, or nonlinguistic states of affairs, no trouble should arise.

11. When no component event is apt to occur, we may ignore the distinction between the explanandum and the explanandum event. It is essential to realize, however, that an explanation must not purport to explain the explanandum event in all of its richness and full particularity; rather, it explains just those aspects of the explanandum event that are mentioned in the explanandum.


13. Ibid. See also "Inductive Inconsistencies," Synthese, XII, no. 4 (Dec. 1960).


15. This condition has sometimes been weakened to the requirement that the premises be highly confirmed. I prefer the stronger requirement, but nothing very important hangs on the choice. See n. 76.

16. A premise occurs essentially in an argument if that argument would cease to be (deductively or inductively) correct upon deletion of that premise. Essential occurrence does not mean that the argument could not be made logically correct again by replacing the premise in question with another premise. Essential occurrence means that the premise plays a part in the argument as given; it does not just stand there contributing nothing to the logical correctness of the argument.

17. The requirement of total evidence demands that there should be no additional statements among our available stock of statements of evidence that would change the degree to which the conclusion is supported by the argument if they were added to the argument as premises. See Carnap, Logical Foundations of Probability, sec. 45B.


35. See Salmon, *The Foundations of Scientific Inference*, pp. 83-96, for fuller explication. Note that, contrary to frequent usage, the expression "P(A | B)" is not the probability of A, given B. This is Reichenbach's notation.


40. Also, of course, there are cases in which it would be possible in principle to make a relevant partition, but we are playing a game in which the rules prevent it. Such is the case in roulette, where the croupier prohibits additional bets after a certain point in the spin of the wheel. In these cases also we shall speak of practical homogeneity.


43. Ibid.

44. *Aspects of Scientific Explanation*, pp. 399-400. Note that in Hempel's notation "P(G | F)" denotes "the probability of G, given F." This is the reverse of Reichenbach's notation.

45. Ibid., pp. 400-01. In a footnote to this passage Hempel explains why he prefers this requirement to the earlier requirement of total evidence.

46. In working out the technical details of the following definition, it is necessary to recognize that homogeneous partitions are not unique. The transfer of a finite number of elements from one compartment to another will make no difference in any probability. Similarly, an infinite subclass of G, if P(A, C, B) = 0, can fulfill the condition that P(A, C, B) = P(A, B) without rendering A's degree of inhomogeneity > 0. Such problems are inherent in the frequency approach, and I shall not try to deal with them here. In technical treatment one might identify subclasses that differ by measure zero, and in some approaches the conditional probability P(A, B) is not defined if P(A, C) = 0.

47. I am grateful to Prof. Douglas Stewart of the Department of Sociology at the University of Pittsburgh for pointing out that the concept of variance is of key importance in this context.


71. Ibid., secs. 50-51.
72. Salmon, "Who Needs Inductive Acceptance Rules?" Because of this difference with Carnap—i.e., my claim that inductive logic requires rules of acceptance for the purpose of establishing statistical generalizations—I do not have the thoroughgoing "pragmatic" or "instrumentalist" view of science Hempel attributes to Richard Jeffrey and associates with Carnap's general conception of inductive logic. Cf. Hempel, "Deductive-Nomological vs. Statistical Explanation," pp. 158-63.
74. See "Aspects of Scientific Explanation," secs. 3.3-3.4. In the present essay I am not at all concerned with explanations of the type Hempel calls "deductive-statistical." For greater specificity, what I am calling "statistical explanation" might be called "statistical-relevance explanation," or "S-R explanation" as a handy abbreviation to distinguish it from Hempel's D-N, D-S, and I-S types.
78. Hempel, "Deductive-Nomological vs. Statistical Explanation," sec. 13, and "Aspects of Scientific Explanation," sec. 3.6. Here, Hempel says, "Nonconjunctiveness presents itself as an inevitable aspect of [inductive-statistical explanation], and thus as one of the fundamental characteristics that set I-S explanation apart from its deductive counterparts."
80. Reichenbach, *The Direction of Time*, chap. IV.
81. The hints are provided in sec. 3.