
**METHODOLOGY
OF SCIENTIFIC SEARCH**

Causal Criteria in Medical and Biological Disciplines: History, Essence, and Radiation Aspect. Report 1. Problem Statement, Conception of Causes and Causation, False Associations

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Abstract—Report 1 of a three-part review examines the conceptual formulation and relevance of the problem, including weak penetration of specific methodologies for causation proving into experimental and descriptive disciplines that study the effects of the radiation factor on living organisms. The philosophical and scientific concepts necessary for understanding the meaning, essence, and possibility of practical application of the criteria (rules, principles) for establishing the truth of associations revealed in medical and biological disciplines are presented. Five types of definitions of the causes and causality were found, which vary from the simplest explanatory (“by production”) to the complex, for deterministic and stochastic effects (necessary and sufficient causes, component causes, probabilistic causes, and counterfactual causes). Many of these definitions originate from famous philosophers (mostly D. Hume). A selection of statements revealing the scientific, practical and social goals of epidemiology and other causality studies important for human life and activity is presented. These goals are primarily related with evidence of the truth of the revealed dependences of effects on agents and impacts, however methods for their achievement can be based on different rules and ethical foundations established on scientific or social tasks. In the second case, the “precautionary principle” is used, and the norms of research developed for application in the scientific community are simplified, in many respects being replaced by prevention or at least reduction of risks, even if the reality of the latter does not have strict scientific evidence. Examples of false but statistically significant associations from various biomedical and social spheres (including estimates of effects of radiation exposure) caused by confound factors are presented. These examples indicate the need to use standardized criteria for assessing the truth of causality.

Keywords: philosophy of natural sciences, biomedical disciplines, causes of effects, causality, confounding factors

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INTRODUCTION

The essence of the biomedical concept of causality is not well known and is rarely considered in experimental and descriptive disciplines related to the effects of ionizing and nonionizing radiation (radiation biophysics, radiobiology, radiation medicine, radiation epidemiology, radiation hygiene, etc.). In addition to the extensive experience of the author of this study, this is also evidenced by the material of voluminous domestic specialized manuals on listed radiation disciplines. Of note is the small number of sources identified through PubMed using specific thematic keyword combinations (for example, “Bradford-Hill criteria & radiation” or “Hill’s criteria & radiation”). The basic epidemiological concept of causality and the criteria for the truth of statistical associations remains little in demand in areas that could have a cardinal effect even on the very existence of society (due to the spread of the radiation factor during peacetime and,

probably, during wartime). Moreover, it is apparently little known to most researchers (at least in Russia).

The presented review that consists of three reports made an attempt to fill the indicated gap. Report 1 considers the general concept that determined the formulation of the problem, well-known philosophical and scientific definitions of causes and causality, and scientific and social goals of research, while listing examples of false associations due to the effect of confound factors.

CONCEPTUAL FORMULATION OF THE PROBLEM

The ultimate goal of descriptive and experimental natural science disciplines is to confirm the existence of certain events and/or to establish a relationship with the causes that have produced them [1, 2]. According to the philosophy of science of this plan, the task of biomedical areas is to identify the causes of phenom-

ena (cases) such as a disease [2], or, in the general sense, health-related states or events in populations [3, 4]. The principle of causality seems obvious in everyday terms; it is the basis of the first steps of science (*de nihilo nihil*—“nothing comes from nothing”)¹ [5] (the list of notes follows the main text).

However, what is an event, or a case? It would seem that the question is strange, and it is almost not considered in literature on the causality of biomedical effects (‘health effects’). According to an explanatory dictionary, an “event” or “case” is what has happened, taken place; it is a fact [6]. Let us try to imagine different events in medicine and biology. On the one hand, an event and fact is, for example, a case of an infectious disease or radiation burn in a particular individual who has undergone corresponding exposure. A case is also a certain change at the molecular/cellular level if the magnitude of the impact exceeds a certain threshold. In other words, a case in this respect is what has happened to a specific and discrete biological object. On the other hand, if we take a population of people exposed to the effect of, for example, radiation or chemical agents in significant doses, we can observe an increase in the incidence of cancer, but at the same time cases of cancer do not necessarily occur in specific individuals [2–5]. It is also possible to irradiate a cell culture in a very large dose, observe the death of most cells, but still there will be individual cells that will survive due to inter-screening and other unpredictable effects [7]. As a result, as for the human population, it will never be possible to say exactly which cells will die and which ones will survive. Is this an event that is equivalent to those mentioned above? Since it does not necessarily happen to an observable single object under *any conditions*? Probably yes, since it is also a fact, i.e., a consequence of an already stochastic effect. The causes for those and other events may turn out to be of different nature. According to, for example, the study [8], they may be of deterministic and probabilistic nature (randomness is also assumed).

Another logical question arises: *what* can be called the cause of an event? Again, it seems that everything is clear from the ordinary point of view and, at the same time, nothing is clear. According to the mentioned explanatory dictionary, “a cause is a phenomenon that produces, determines the appearance of another phenomenon” [6]. It is noted in [9] with reference to [10] that “... most researchers would find it difficult to define the words in any but a circular fashion; causes are conditions and events that produce effects, and effects are conditions and events produced by causes.”²

In previous centuries, famous philosophers nevertheless tried to determine the cause and causality more accurately [2, 5, 11]. The results were not unambiguous (more details will be given below). A special study of 2001 [8] on sources with definitions for the period

from 1966 demonstrated, first, that the term “causality” appeared as a MeSH term in PubMed only in 1990 (although a similar term ‘causation’ can be already found in scientific sources of the 19th century [12]). Second, there are as many as five categories of causality [5, 8] (they will be discussed below). However, as indicated in [5], the definitions found in literature alone are not enough to provide a basis for understanding the causes of pathology. In a hitherto program (for example, [2, 4, 5, 10, 12]) publication on the criteria of causality that was published in 1965, its author, English statistician Sir Austin Bradford Hill (1897–1991) immediately said, “I have no wish, nor the skill, to embark upon a philosophical discussion of the meaning of ‘causation’” [13].³

Another founder of causality principles, Mervin Wilfred Susser (1921–2014; the United States),⁴ defined the cause as “... any factor, whether event, characteristic, or other definable entity, so long as it brings about change for better or worse in a health condition” [17] (cited from [9]).⁵ Later, by 1991 [11], M. Susser rethought the definition of a cause in a pragmatic sense as simply “... something that makes a difference.”⁶ Thus, everything returned to the level of the explanatory dictionary.

The next question is as follows: is it always clear where the cause is and where the effect is? Is it always obvious “what arose earlier—the chicken or the egg?” At first glance, this is just sophism, however the situation changes if we turn to the phenomenon of “reverse causation” [18]. When a phenomenon that appears to be the result of a certain cause is not actually connected with it, but is connected with *its own* previous sources that have determined this very “cause” (this definition is ours).⁷ Although there are enough examples, it is clear that the relevant examples are those associated with the radiation factor. More than a dozen epidemiological studies conducted over the past five to six years have shown an increase in the risk of cancer and/or leucoses after computed tomography (CT; see a list of sources in [19, 20]). However, the analysis of the specifics carried out by the main (apparently) US radiation epidemiologist, John Boyce (John Boice, Jr.), led to the conclusion on the leading role of the confound factor of “reverse causation.” CT was more often performed in those who were suspected to have malignant neoplasms [19]. This viewpoint is also shared by the UN UNSCEAR [21, 22]. Similarly, the cases of thyroid cancer after the therapeutic effects of radioiodine are attributed, not to the radiation factor, but to those initial no-cancerous pathologies that caused this therapy (since hyperthyroid states, benign neoplasms, etc., increase the risk of thyroid cancer) [23].

Finally, how can we not only simply determine as strictly and scientifically as possible, but also prove that something (an event, factor, impact) was the cause of one or another phenomenon, which is

important in the biological, epidemiological, medical, environmental, or social respects [2, 5, 8–11]? The common scientific idea implies that one should identify the association, establish the magnitude and statistical significance of correlation, and also show the presence of a dose–effect relationship (rarely something more). The world of experimental biology and medicine and descriptive and statistical epidemiology (together with sociology, which contributes to epidemiology [2, 9, 11, 17]) is a world of associations and correlations; they fill scientific journals, often as the last word of evidence. This problem concerns molecular disciplines and population studies. It is unlikely to be a mistake to say that the vast majority of researchers in the field of biomedical effects of various exposures (including radiation) imagine the scientific evidence of causality just in this way. This applies, not only to Russia and neighboring countries, but also to a significant part of the rest of the world, which was shown, in particular, by mass alarmist conclusions based only on identified associations in the period after the Chernobyl nuclear power plant accident [24, 25].

Meanwhile, association does not mean an obligatory “causation” (causality) [13], and everyone has a spontaneous idea about it, however this often fails to help when the matter concerns scientific research. The sun does not rise because the rooster has crowed. The frequency of deaths from drowning [26, 27] or from killing [28] is in no way related to the level of ice cream consumption, no matter how strict, statistically significant and distinct correlations are, perhaps, even with the presence of a “dose dependence” [26–28]. The male sex does not determine mortality from lung cancer, no matter what statistics evidence this [5, 29].⁸ And there may be more than a dozen associations of this type (a number of them are discussed below).

In addition to philosophers [2, 5, 11], researchers in the field of biomedical effects were fully aware of the indicated problem since a recent period (let us recall the establishment of a “causal relationship” in the above-cited publication of the 19th century [12]). A set of simple criteria for establishing the causation of then particularly significant infectious diseases appeared (Henle–Koch criteria or postulates; F.G. Jacob Henle, R. Koch, 1877, 1882), which did not prove to be absolute [2, 9, 16, 32–34]. As for noninfectious pathologies, especially stochastically induced, the centuries of development of medicine and biology in this area passed in the dark: the generally accepted approaches to establishing the truth of associations were absent until the 1950s–early 1960s, when, at long last, strong evidence of at least harm of smoking was obtained.

Guesses about the possibility of smoking leading to lung cancer appeared in the late 19th century; in 1912, they took shape in the specific assumption about smokers (Isaac Adler; the United States), which strengthened in the 1920s–1940s, when the first evi-

dence-based works were carried out using the case-control method, although they were still imperfect (1939 and 1943; Germany). By the early 1960s, all doubts had disappeared: dozens of retrospective and prospective studies all over the world had already been accumulated [35].⁹ This led to realizing the need for certain rules, criteria, or some kind of guiding principles, with the help of which it would be possible to confirm the truth and plausibility of detected associations for noninfectious pathologies. In the 1950s, a number of authors formulated individual principles-criteria [11]¹⁰; five of them were included in the program report of the Surgeon General (Chief Physician [33]) of the USA with the accompanying document of 1964 on the consequences of smoking [43].¹¹ Then, in 1965, the already mentioned brief program publication of A.B. Hill appeared [13], where there were already nine criteria (the known “Hill criteria” or “Bradford–Hill criteria” [2, 5, 11, 15, 16, 29, 32–34, 36–42, 44–46]). These criteria (or “guidelines”) are planned to be analyzed in the following reports; here we say that Hill added only one criterion, and the rest can be found in the works of earlier authors.

The first above-mentioned *monograph* on the criteria of causality [17] was published in 1973; author M. Susser claimed to have developed these criteria independently [11]. An exhaustive set of 10 criteria (later they began to be called “postulates” [15, 16]) of causality for pathologies of *all* types appeared in 1976 (‘unified concept’ [34]). These postulates were formulated by Alfred Evans (Alfred Spring Evans; 1917–1996; the United States) [34, 45].

Since then, as stated in [9], “these guidelines have generated a talmudic literature on their nature, logic, and application.”¹² Our study has fully confirmed this statement. One can find sources listing these criteria (but without much analysis, i.e., not “talmudically”) even in the Russian-language literature, the translated and the original ([32, 33] and relatively massively in RuNet. For many decades that have passed since the 1960s, the clarity of using the criteria has somewhat faded in connection with the attempts to narrowly regulate and specify various concepts, as well as sometimes scholastic criticism of specific issues, etc. (for example, [2, 5, 9–11, 17, 29, 34, 36, 38–40, 44–46]).

Meanwhile, it is obvious that research in the field of biomedical problems without making allowance for the principles of causality are almost blind research, whether these be research on radiation effects at any level or effects of other factors. A molecular radiobiologist reveals certain effects and associations and believes that he has received evidence that is important for medicine. A researcher of *in vitro* effects, setting up experiments, also considers them evidence-based and extends them to human health, rarely thinking about the obvious abnormality of almost any long-lasting cell cultures.¹³ However, what is the potential of transferring the revealed patterns even from animals to

humans, not to mention experiments at the molecular-cellular level? [47].¹⁴ A specialist in the field of clinical medicine once again considers his own observations to be final and suitable for practice, but does he have a sufficient sample and adequate “controls” to draw conclusions about the true cause of the discovered relationships? Finally, an epidemiologist, looking at them from above, believes that only his area makes it possible to obtain data for calculating risks [8, 11, 17, 45] and for using them within, for example, radiation safety [50, 51]. However, from a formal point of view, an epidemiologist only identifies associations that are similar to the aforementioned relationship between ice cream consumption and mortality from drowning or killing.

At what levels of evidence of causality is the activity of these specialists? Are they all needed and can they replace one another here? At one time, we tried to clarify this issue [50, 51], although not in full. By all indications, the concept of causality criteria and principles of evidence continues to be almost absent in the Russian scientific area that studies the effects of the radiation factor. Apparently, it is absent in other near-medical areas of Russia, despite the analysis of the Hill criteria [13] and the corresponding developments of K. Rothman (K. Rothman; the United States) [10, 44] in a manual on epidemiology (2006) [33] and that they are widely represented in RuNet.¹⁵

One review, even if it consists of three reports, cannot cover all available literature within this topic over the past more than half a century. However, as we hope, our study has covered the main sources (including most of the basic originals) and, most importantly, the provisions worked out over the indicated period together with the cardinal issues of the historical and critical plan. The presented cycle of reports has the task of reproducing the main provisions and significant stages in the formation of the principles of causality of effects, their essence, limitations, and the possibility and breadth of application. Although the actual part (examples) is often within the radiation field, the conceptual importance of information goes far beyond its borders.

HISTORICAL AND PHILOSOPHICAL FOUNDATIONS OF THE CONCEPT OF CAUSES AND CAUSALITY

The very concept of “causality” and the developed criteria (principles) of establishing causal relationships in biomedical disciplines go back to the philosophical developments of past centuries. Therefore, almost all relevant sources that consider the theory of causality mention and even consider the provisions of a number of philosophers (Table 1).

From Table 1²⁴ the greatest contribution to the concept of causality in the field of biomedical disciplines was made by the ideas of a subjective idealist,

agnostic D. Hume, on whose provisions the empirio-criticism of E. Mach and R. Avenarius was once based [71].²⁵ D. Hume is mentioned in 17 of 22 specialized sources (Table 1). Second place is taken by J.S. Mill (seven of 22 sources) who is more inclined to the recognition of reality, including on the basis of experience.

On the whole, when getting acquainted with philosophical and natural science literature on causality, one gets the impression of the complete domination of David Hume’s ideas, which are notable for an unprecedented breadth of coverage for all eras [59]. Some provisions of D. Hume are considered in almost every specialized source.

At first glance, it seems that similarity with the provisions of the subjective idealist, moreover, the one who denies any reality in the nature of causal relationships, is not a very good position in developing scientific rules for establishing these relationships. However, as in the still memorable development of dialectical materialism that is based on, in particular, Hegel’s idealistic provisions, the success in this respect has also been achieved [38].

We, just as Sir Austin B. Hill in 1965 [3] (see the previous section), do not have only the “wish or skill,” but also a sufficient space in this review for an in-depth historical and philosophical discussion of the concept of “causality.” In addition, as noted in [8], philosophers and epidemiologists have different objects of study. While philosophers develop general principles of causality by seeking definitions, epidemiologists are usually interested in specific examples of causal relationships, building causal models [8].

Therefore, it is proposed not to go beyond the information on the mentioned topic, which is presented in Table 1, and the corresponding “Hume Rules” and other relevant points from philosophical ideas will be described below, with reference to modern criteria (Report 2). Here, we will consider only the main publications, which analyze the concept of “causality” in natural sciences.

CONCEPTS OF CAUSALITY IN MEDICAL AND BIOLOGICAL DISCIPLINES

Making a preamble to the section, we cannot but mention statements such as “Physics and chemistry have completely abandoned the concepts of cause and effect. These terms are no longer used in these sciences [since the end of the 19th century]” [5]. This state of affairs has developed in connection with the introduction of the quantum theory, which has led physics to accept indeterminism and to abandon the classical concepts of cause and effect [72, 73].

Although well-known statistician Karl Pearson (1857–1936) rejected the concept of causality as a “secret fetish” as early as 1892 [73] (cited from [72]), the situation remains somewhat different for medicine

Table 1. Philosophers whose provisions on causality/evidence of causality are considered in modern biomedical sources of the topic*

Philosopher	Presented area	Brief information on provisions on the topic	Specialized references to the topic with mentions
Aristotle (384 BC–322 BC); Ancient Greece	Creator of scientific philosophy and physics [52]	Formulator of the “Four causes of being”: (1) material cause (of which all objects consist); (2) formal cause, in which the form manifests itself, making the essence, substance of being; (3) acting or producing cause (what the object does); (4) target or final cause, answering the questions “Why?” and “For what purpose?”	[53, 55]
Francis Bacon (1561–1626); England	Founder of empiricism and natural philosophy [52, 56]	Development of the inductive method in science; experiment [52, 56]. “Deductive logic could never be predictive without the fruits of inductive inference” [57] ¹⁶	[33, 57]
Galileo Galilei (1564–1642); Italy	Creator of classical mechanics [52]	The statement that “true causes” are necessary and sufficient to produce a certain effect. True causes are universal ¹⁷	[8, 11, 55, 58]
Rene Descartes (1596–1650); France	Creator of analytical geometry [52, 54]	Fundamentalist: the method of doubt, or methodological skepticism. Rationalism: knowledge is based on judgment (“I think; therefore I am”) [52, 54]. Causation is only a fictitious construction [55]	[55]
John Locke (1632–1704); England	Empiricism [52, 56]	Popularization of inductive methods of F. Bacon; the statement that empiricism is the basis of scientific philosophy. The position that knowledge is based on the data of subjective meaning towards which experiment gives rise [57]	[53, 55, 57]
Isaac Newton (1643–1727); England	Physicist, mathematician, mechanic and astronomer; philosophy of natural sciences [52, 56]	Newton replaced cause and effect with functional relationships; he introduced the provision on the minimum sufficient cause [5]	[5, 8]
George Berkeley (1685–1753); Ireland	Associate of empiricists. Theory of perception [52, 54–56]	Berkeley developed the thesis that “being is either that which is perceived, or the one who perceives” [54]. Causation can only be an attitude — a fictional mental construct [55]	[55]

Table 1. (Contd.)

Philosopher	Presented area	Brief information on provisions on the topic	Specialized references to the topic with mentions
David Hume (1711–1776); Scotland	Subjective idealism, agnosticism, empiricism [52, 56–61]	Causation is only a regular association between two events occurring sequentially. Denial that causation can be proven; provision on the subjectivity of knowledge and falseness of inductive reasoning (inductive logic cannot establish a fundamental connection between cause and effect). Wording of the “cause and effect rules by which to judge” ¹⁹ [2, 11, 38, 44, 57–61]	[2, 5, 8, 9, 11, 14, 33, 37, 38, 40–42, 44, 53, 55, 56, 58]
Immanuel Kant (1724–1804); Germany	Founder of German classical philosophy [11, 52]; he is sometimes included among agnostics [60]	Kant considered the physics of Newton to be the eternal truth [5]. Space and time are subjective, but they are of sensory contemplation, i.e., those given to a person before experience [60]. “Reasoning was to imply the principle of causality from the assumption that it is among the conditions of every experience” [5]	[5, 11, 55]
Pierre-Simon de Laplace (1749–1827); France	Mathematician, mechanic, physicist and astronomer [52, 54]	Scientific determinism: everything in the physical world can be explained using the laws of mechanics. The concept of the causal explanation of the evolution and change of large systems (Laplace determinism) [52, 54]	[55]
John Stuart Mill (1806–1873); England	Empiricist, sociologist, economist, politician [52, 54–57]	Mill developed Hume’s causal logic [44, 53, 57]. He formulated five inductive canons to prove the reason for the connection of observed associations: canons of similarity, difference, combined canon of similarity and difference, canons of residues and related changes [11, 53, 62] ²⁰	[9, 11, 44, 53, 55, 57, 63]
Ernst Mach (1838–1916); Austria	Positivism; subjective idealism; empirio-criticism [52, 56, 60]	“... alluding to Hume, Mach stressed the psychological nature of the concepts of causality and pointed out that ‘in nature there is no cause and no effect’ and that these concepts are results of an economical processing of perceptions by the human mind” [5] ²¹	[5]

Table 1. (Contd.)

Philosopher	Presented area	Brief information on provisions on the topic	Specialized references to the topic with mentions
Bertrand Arthur William Russell (1872–1970); England	Philosophy of science, a mathematician and public figure. Neopositivism (logical atomism) [52]	Russell rejected the concept of causality [8]: “that causation “is a relic from a bygone age, surviving, like the monarchy, only because it is erroneously supposed to do no harm” (B. Russell, 1959; cited from [2]) ²² He refuted Hume’s subjective view of the world by the demonstration, which he attributed to Immanuel Kant, that knowledge exists independent of experience. Russell showed that relationships, too, can exist independent of experience [11] ²³	[2, 8, 11]
Karl Raimund Popper (1902–1994); Austria, England	Modern philosophy of natural sciences; criteria of scientific nature of hypotheses [52]	Application of K. Popper’s criterion of falsifiability of hypotheses in epidemiology [11, 64–67], in particular, in checking the significance of causality criteria [11, 66]	[11, 33, 36, 56, 66, 67]

* We emphasize that Table 1 mentions and analyzes the provisions of philosophers *only* in epidemiological and biomedical sources. The philosophical question of causality throughout the world is certainly broader, and this was, to a varying degree, considered by almost all generations of thinkers. For example, the accessible first chapter of the monograph *From Cause to Causation* (2002) [68] examines the history of the concept of causality, starting with Ancient Greece (Aristotle, the Stoics) and the Middle Ages (Thomas Aquinas). The review considered the provisions of Descartes, Hobbes, Spinoza, and Leibniz; the critical philosophy of Locke; and the corresponding constructions of Newton, Hume, Kant, and Mill. The author [68] notes a certain evolutionary development of the concept of causality.

and biology. So, in 1934, no less famous statistician Robert Fisher (1890–1962) noted that uncertainty must not be incompatible with causality; simply an action (cause) influences the corresponding probabilities “as if it predetermined some of them, excluding others” (cited from [72]).

As already mentioned in the previous section, biomedical and, specifically, epidemiological sources present no single formulation of the concepts of “cause” and “causality.” The study by M. Parascandola and D.L. Weed, 2001 [8] presented a literature review to identify and systematize the definitions used. These definitions fit into five categories having strengths and weaknesses depending on the situation, the scope and type of effects (pathologies). It was noted [8] that the scientific definition of causality should, on the one hand, make it possible to differentiate causality from simple correlation (association), and, on the other hand, it should not be so narrow as to exclude obvious causal phenomena from consideration. M. Kundi, 2006; 2007, studies [5, 29] supplemented and discussed these approaches from [8]. The author of [5, 29] cited the data from [8] in his publications with additions and remarks. Table 2 presents the combined information from [5, 8, 29].

A total of five definition categories that are given in Table 2 correspond to the main approaches to causality in philosophical literature. Although these definitions are not mutually exclusive, there is a fundamental difference between deterministic and probabilistic concepts of causality, which can lead to consequences in preferring some definitions over others. So, if the philosophical deterministic definition of causality is taken as the basis, only deterministic models will be recognized as causal in epidemiology. Hence, it is important that epidemiologists understand the practical consequences of adopting a specific type of definition of causality [8].

All categories of concepts are not absolute and are not notable for completeness, as can be seen from the material below.

Cause “by Production”

According to D. Hume, “a cause is when one object (event) follows another, and when all objects similar to the first are accompanied by objects similar to the second” (the “strong concept”) [9, 40, 59, 61]. The “working” definition of M. Susser has already been already mentioned above [11]: “... something that makes a difference” (see note 6). Thus, a cause is

Table 2. Definitions of the term “causation” in epidemiological literature (according to M. Parascandola and D.L. Weed, 2001 [8]; M. Kundi, 2006, 2007 [5, 29])²⁶

Category defined according to [8]	Essence according to [8]	Essence according to [5, 29]	Brief critics [5, 29]
“By production”	Causes are the conditions that play a significant role in the occurrence of a disease	A cause is what produces or creates an effect. A cause is a condition, under which an effect arises	Tautology: “production” and “creation” are synonyms of “causation”
“Necessary causes”	A necessary cause is a condition, without which an effect cannot occur	A cause is a condition, without which an effect cannot occur	Only a few diseases can have a specific single cause
“Sufficient-component causes”*	A sufficient cause ensures that an effect will occur when it is present. A sufficient-component cause* consists of several sufficient components, none of which individually is sufficient, but together they structure a sufficient cause	A cause consists of several components, none of which is sufficient in itself, but together they must lead to an effect	This introduces excessive complexity in constructing the dose dependence, and in the interaction of components
“Probabilistic cause”	A probabilistic cause increases the likelihood of an effect. It is neither necessary nor sufficient	A cause is a condition that increases the likelihood of an effect	A distinction between association and causation is not drawn
“Definition of counterfactual cause”; “counterfactual cause”**	A counterfactual cause influences an effect (or the probability of the effect) when it is present in comparison with its absence, while all other conditions remain constant. This cause is neither necessary nor sufficient	A cause is the condition under which there is a difference (probability of a difference) in an effect	This is, in the strictest sense, unprovable, since there is only one world, and no one can observe the same event twice with a specific condition (“cause”) and without it ***

* According to the manual on epidemiology by V.V. Vlasov, 2006 [33], the translation in these cases is expressed with the terms “sufficient” and “component cause.” The generally accepted analogue “Sufficient-component causes” failed to be found in Russian-language medical and biological literature.

** The main translation was taken from the *Dictionary of Epidemiology* edited by J.M.A. Last (counterfactual definition) [15]. Well-known English-Russian dictionaries do not include this word. “Counterfeit” is a corresponding Russian-language term on the Internet (in particular, in Google Translator), however this is probably modern slang calque. We use the word “counterfeit,” which, at least, is found in publications on the history and theory of language.

*** This aspect has received a scholastic-philosophical discussion in more than one source. The following is only the simplest example: it is not possible to observe the same person (“under otherwise equal conditions”) who both has never smoked and has smoked for years to compare and reveal the effect of just and only smoking [8, 53].

what produces, makes, or creates something, or leads to something, “leads to a result(s)” [8].

The foregoing shows an ontological difference between causal and noncausal associations. However, there is no clarity as to what “production” or “creation” means. Causality is defined in terms of another equally elusive concept [8, 75]. It is in connection with this that philosophers such as D. Hume [2, 11, 38, 44, 57–59, 61] and B. Russell [2, 8, 11] rejected the concept of causality. The defectiveness of this definition was noted in epidemiological sources [8, 75].

Necessary and Sufficient Causes

A necessary cause is a condition without which an effect *cannot* occur, and a sufficient cause is a condition under which an effect *must* occur [2, 5, 8, 10, 11, 29, 32, 33, 41, 44] (a “sufficient cause is a cause that *inevitably* produces an effect” [10]). Four types of causal relationships are derived from the following:

- (1) necessary and sufficient,
- (2) necessary, but insufficient,
- (3) sufficient, but not necessary, and
- (4) neither necessary nor sufficient [2, 8, 33] (Table 3 and Fig. 1).

Table 3. Explanatory diagram of the four types of causal relationships (based on the materials [33])

Necessary cause		Sufficient cause		Necessary and sufficient cause		Component cause	
cause	effect	cause	effect	cause	effect	cause	effect
No	No	No	Yes	No	No	No	Yes
Yes	Yes	Yes	Yes	Yes	Yes	There is cause 1	No
Yes	No	Yes	Yes	Yes	Yes	There is cause 1+ cause 2+	Yes
An effect does not exist without a cause, although it may not exist even with a cause		An effect always exists if there is a cause, but can also occur without it		An effect is not observed without a cause; it always exists with a cause		Cause 1 produces effects only if there is an additional (complementary) cause (causes) 2, 3.... But an effect does not require all causes	

Discussions concerning what types of cause-and-effect relationships should receive an advantage in epidemiology were and are conducted, and, according to [8], there is even an opinion that the term “cause” should be limited by specific necessary conditions (the sources of this approach are the microbial theory of diseases). Some authors have also extended these patterns to noninfectious diseases. So, the W.E. Stehens publication in 1985 [78] states that each pathology has its own specific, unique cause, and multicausal models simply indicate gaps in scientific knowledge, the fact that the desired cause has not yet been disclosed (some confirmation of this approach is, for example, the discovery in the early 1980s of the etiology of gastric ulcer caused by *Helicobacter pylori* infection [79]). B.G. Charlton noted in his study published in 1996 that epidemiology, like other disciplines, must search for the only necessary cause for an effect to be considered a scientific discipline (“... risk factor epidemiology cannot be considered as a scientific discipline because it aims at concrete usefulness rather than abstract truthfulness”) [80].²⁷

Of course, this strict determinism that requires an unambiguous correspondence between cause and effect, when the same cause must invariably lead to a similar effect, without the role of chance, is not applicable to the field of radiation effects. It follows from the summary of sources given in [8] that the causes of complex chronic diseases such as cancer and cardiovascular pathologies usually do not fall under the type of either necessary or sufficient causality.

Component Causes

The general idea of necessary and sufficient components of a single cause of a disease was spontaneously developed many decades ago (see the I.V. Davydovskii, 1962, study [74]), however the wording of the definition of causality on this basis is

associated with the name of Kenneth Rothman, whom has already been mentioned in the previous section (from the first publication on the topic in 1976 [10] to a number of fundamental subsequent studies up to 2008 [44, 82–84]²⁸). The model of this author, moreover, with mandatory personalization, is now analyzed/presented in various English-language [8, 11, 15, 46, 57, 58, 72, 86–94] and Russian-language sources (including Runet) [33, 95, 96].²⁹ Apparently, Kenneth Rothman is currently one of the main authorities in establishing the rules for the multicausality of effects in epidemiology, although his main provisions do not seem so original; they are perhaps more detailed from the epidemiological standpoint. The manual on epidemiology (V.V. Vlasov, 2006 [33]) says: “Modern epidemiology uses a causality model called a component. Its most developed version belongs to K. Rothman.”

The idea of the totality and interaction of various causes for a single effect was put forward a little earlier in the work of D. Lewis, 1973 [97] (in philosophical aspect). In 1974, a philosopher J.L. Mackie formulated the now well-known [2, 8, 41, 57] concept of the condition INUS (‘Insufficient and Non-redundant part of an Unnecessary but Sufficient condition’) in the monograph [98].³⁰ In addition, as noted in [99], J.L. Mackie first developed a model of component causality in the same 1974 [98], i.e., before K. Rothman.

Finally, it was indicated [8, 68] that the sources of the concept of multicausal phenomena go back to John Mill’s philosophy [62] (the provision that the real cause of an effect (a sufficient condition) is a combination of a number of conditions (that are necessary, but not sufficient if taken separately) [68]).

Figure 2a presents our conditional modification³¹ of Rothman’s “Causal Pie Model” [10, 44, 82].

The concept is based on the fact that most of the causes that are important in the medical and epidemi-

Necessary and sufficient causes
of “Table 2 × 2”

(*a, b, c, d*—the frequency of the effect; conditionally)

Necessary cause

	+Tuberculosis pathogen	-Tuberculosis pathogen	(a)
+Tuberculosis	<i>a</i>	0	
-Tuberculosis	<i>c</i>	<i>d</i>	

Sufficient cause

	+ ¹³¹ I, 200 Gy per ThG*	- ¹³¹ I	(b)
+Ablation of ThG (hyperthyroidism)	<i>a</i>	<i>b</i> **	
-Ablation of ThG (hyperthyroidism)	0	<i>d</i>	

* ThG is the thyroid gland
** The surgical way is possible

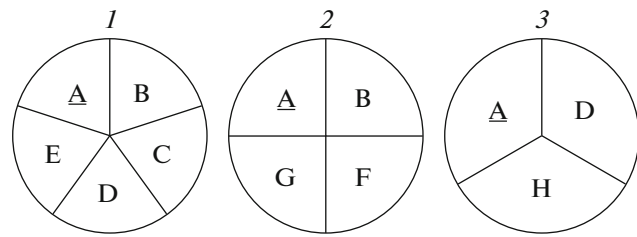
Necessary and sufficient cause

	+Irradiation of the skin (5–10 Gy)	-Irradiation of the skin	(c)
+Radiation burn	<i>a</i>	0	
-Radiation burn	0	<i>d</i>	

Fig. 1. Schemes of necessary and sufficient reasons in “2 × 2 tables”. The general meaning is presented in [36, 64]. The example (a) is widespread in the sources on the causality of biomedical effects [2, 8, 10, 15, 33, 41] (etc.); (b) and (c) are our examples. In ablation of thyroid tissue in hyperthyroid states (b), therapeutic administration of ¹³¹I in an appropriate dose (according to [76], the necessary dose) is a sufficient cause, however a similar effect can also be achieved via surgical removal of organ tissues [76]. Local exposure of the skin, for example, exposure to β-radiation in doses of 5–10 Gy leads to radiation burns (c), which have specificity in their development, therapy, and prognosis, differing from other types of burns [7, 21, 77]. That is, when a threshold absorbed dose is reached, radiation burns *always* occur, and it will not be possible to completely replace the radiation exposure with another physical or chemical factor that also causes burns.

ological aspect are components that constitute sufficient causes, however they are not sufficient in themselves. K. Rothman mentions the “constellation” of phenomena (sectors in Fig. 2), which is a sufficient cause (circles in Fig. 2). Moreover, the entire “constellation” and its components are called “causes.” As a result, the term “cause” alone does not indicate whether the phenomenon will be a sufficient cause or a component of a sufficient cause [10, 44].

(a)
Component sufficient causes of effect



(b)

Component sufficient causes of tuberculosis

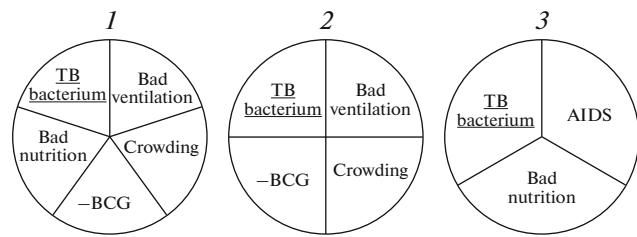


Fig. 2. Model of K. Rothman’s “Causal Pie” (the components of a sufficient component cause of effect): (a) the general scheme (the conditional modification of the original from [10, 44, 82]); (b) in relation to three sufficient causes of tuberculosis (based on the materials [91]).

Developing the concept in a 2005 publication [44], K.J. Rothman and S. Greenland provide figurative examples of what must be understood as component causes. Since the first assessment of the concept of causality is based on direct observations, it is often limited only by their scope. For example, when a light switch is in the “on” position, a light effect is usually observed. However, it is clear that the causal mechanism for generating light includes more components than just turning the switch. Nevertheless, the tendency to identify a certain “switch” as a unique and sole cause follows from its *usual*, observable role as the final factor acting in the causal mechanism. Electrical wiring and other components, starting with the production of electricity at a remote station, can be considered as part of the causal mechanism, but once they are put into operation; so, they rarely require additional attention. Thus, the switch may appear to be the only part of the mechanism that must be activated to obtain the effect of switching on a light. The inadequacy of the assumption of the sufficient role of the switch will manifest when a light bulb burns out or something happens in the power supply [44].

There are many examples of component causes associated with infectious and noninfectious diseases. Just drinking contaminated water is not enough for falling ill with cholera (which Max Pettenkofer partially demonstrated at one time), and smoking is not enough alone to induce lung cancer, however both

factors are *necessary* components of *sufficient* causes, as indicated in [10]. No single component (“sectors”) of the constellation is sufficient, although, as can be seen from Fig. 2a, some component (A) is included in all causes. This is a *necessary* component (cause), such as, for example, the presence of infection with tuberculosis bacterium in all sufficient component causes of this disease (Fig. 2b), just like the presence of infection with another agent in other infectious pathologies, such as the mentioned cholera pathogen.

The following question arises in light of the biomedical effects of radiation: “Can radiation be a necessary component of the “constellation” of a sufficient cause of certain radiation effects? At the level of an organism, or at least a cell?” Apparently, it cannot, because it is known that specificity is absent for radiation effects at these levels. There is no pathology or change that would be induced, not solely by irradiation, but by some other factor (at least reactive oxygen species), just like there is no corresponding radiation biomarker [7, 21]. As already mentioned, in our opinion, some exceptions may be radiation burns, which have differences in their development, therapy, and prognosis from other burns [7, 21, 77] (Fig. 1c); the other exception is probably radiation syndrome. The lack of radiation specificity of the effects is illustrated by the schemes of sufficient component causes of radiogenic cancers (a similar composition can be built for leucoses) and radiation cardiovascular pathologies (Fig. 3).

From the diagrams in Fig. 3, a completely similar pathology, the component cause of which nevertheless does not include irradiation, can be theoretically found for a seemingly radiogenic cancer. The situation is the same for cardiovascular and cerebrovascular pathologies, the attribution of which to radiation exposure is difficult to establish even at doses less than 0.5–1 Gy of radiation with low LET [101–103] (the list of components of causes can be continued in both cases considered).

The identification of all components of one or another sufficient cause is not necessary to prevent an effect, since blocking of the causal role of even one component makes the joint action of other components already insufficient and, therefore, eliminates the entire action [10]. In his original study [10] K. Rothman gives an example of removing the smoking component from a sufficient cause that provokes lung cancer, which can completely neutralize the effect of the entire constellation of causes, many of which cannot be recognized.

The presented model also illustrates the concept such as “*strong*” and “*weak*” component causes. A component cause that can complete the whole complex of a sufficient cause (that is, 1–3 in Fig. 2), but only in combination with other components with a low prevalence, is a “*weak* cause.” The presence of this component cause changes the probability of an out-

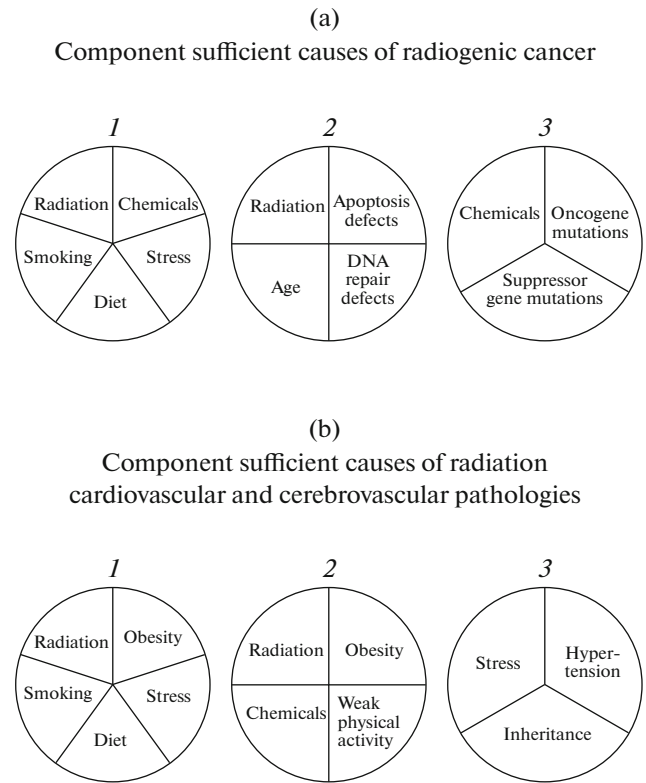


Fig. 3. Component sufficient causes of radiogenic cancer (leucosis) (a); and radiation cardiovascular and cerebrovascular pathologies (b). The schemes were made taking account of information from [77, 87, 91, 100].

come insignificantly, from zero to a value slightly more than zero, which reflects the rarity of additional components. However, if these additional causes are not rare, but are almost ubiquitous, then this makes “*strong*” the component cause that completes the entire causal complex together with them. A weak cause brings only a small increase in the risk of a disease, while a strong cause will significantly increase the risk [10].³²

Thus, the ‘strength of a causal risk factor’ depends on the prevalence of additional components in the same sufficient cause. However, this prevalence often depends on customs, circumstances, or accidents, not being a scientifically based characteristic [10]. At this point, K. Rothman considers the corresponding example of a component sufficient cause from the study by B. MacMahon, 1968 [104], which is related to the phenylketonuria disease. In a population where most people have a diet with a high content of phenylalanine, the inheritance of the rare phenylketonuria gene is a “*strong*” risk factor for mental retardation, and phenylalanine in the diet will be a low risk factor. However, in another population in which the phenylketonuria gene is very common, but at the same time few people have a diet with a high level of phenylalanine, gene inheritance will be a low risk factor, how-

ever phenylalanine in the diet will be a strong risk factor. Thus, the strength of a causal risk factor for a relative risk parameter depends on the distribution of other causal factors in the population within the same sufficient cause. The term “strength of a causal risk factor” has meaning as a description of the importance of this indicator for healthcare, although it is devoid of biological meaning in describing the etiology of the disease [10, 44].

In relation to the sufficient component causes, which include the radiation factor, many situations similar to those described for phenylketonuria can be proposed [10, 104]. For example, when radioactive iodine fell out in the early period after the Chernobyl accident, the component sufficient cause of thyroid cancer could include, among other things, an individual’s age and a diet with stable iodine as components. As a result, the component associated with the radioiodine dose affecting the thyroid gland could be a weak or strong causative factor, depending on the prevalence of the mentioned additional causes. In a low prevalence of normal iodine supply, this could be a strong factor, and, in a high prevalence, it could be a weak factor; the same is true for an individual’s age.

Another example is some hereditary pathology associated with DNA repair defects [105]. When there is a significant level of radiation exposure on the population, for example, in a high natural radiation background and/or medical care that involves many X-ray diagnostic procedures (an additional cause), the causative factor of the repair defect will be strong, otherwise it will be weak.

Finally, based on his model, K. Rothman considers the process of synergy, i.e., the interaction of causal components. Synergy means that two component causes are parts of the same sufficient cause [10]. Although the indicated author is of the opinion that synergistic components do not have any effect separately [10], nevertheless, many analogies can be found for the field of radiation exposure. For example, the UNSCEAR-2000 (Appendix N) that is devoted to the combined action of radiation and other agents [106] reflects the synergistic interactions of radiation with chemicals, smoking, asbestos exposure, etc. There are also recent domestic monographs on this subject (V.I. Legeza et al., 2015 [107], V.G. Petin, J.K. Kim, 2016 [108]).

In light of the foregoing, of particular relevance is the provision of K. Rothman, according to which small doses of an agent in comparison with large doses may require a more complex set of additional components to complete a sufficient cause [10]. In 1988, K. Rothman et al. [82] proposed epidemiological approaches to enhance weak associations. The first approach is to limit the study population to a subgroup of individuals with deliberately low background risk levels, and the second is to reduce an undifferentiated erroneous classification (increasing specificity for

measuring exposure, classifying pathology and length of the latency period).³³

In recent years, the use of the “Causal Pie” model has been moving from theory to practice, not only in medicine and epidemiology [58], although at the beginning of 2018 there were only 12 sources found in PubMed in searching for ‘Causal Pie,’ if we take studies specifically on the topic (starting with 2011); five of them are theoretical developments.

We know the following examples (not only from PubMed).

- Risk assessment for myocardial infarction in the ‘European Prospective Investigation into Cancer and Nutrition-Potsdam Study’ [86].

- A study of the dependence of the frequency of multiple sclerosis on lifestyle factors and the environment [109].

- Analysis of cervical cancer risk in papilloma virus infection [110].

- Assessment of liver cancer risk factors (hepatitis viruses; alcohol) [111].

- Finding the genetic determinants of hypertension at young age [112].

- Risk of bipolar disorders depending on heritable factors and family conditions [113].

- Determination of tick-borne causation of Rosacea [114].

- Construction of sufficient component causes of oral cavity cancer [92].

- The same task for flu (H1N1 virus) [115].

- The same task for tuberculosis [91] (Fig. 2b).

- Use of the “Causal Pie” in evolutionary biology and ecology [88].

Of the listed works, five are Chinese (including Taiwan) and one is Korean (i.e., more than half). The list can be supplemented with our developments for radiation effects (Fig. 3).

Thus, although the cases of direct application of the model developed as early as 1976 [10] are clear to be available, they are few. Perhaps, this is due to some triviality of the idea, which leads to the failure to mention the specific term with ‘pie’. Nevertheless, the similar strategy “according to Rothman” was called “powerful” [58].

To conclude the subsection, we should briefly consider the “Web of Causation” (or “Web of Causal Determination” according to the translation [15]), with which K. Rothman’s scheme is somewhat similar in idea. Since multicausality is the main canon of modern epidemiology [5, 8–11, 13–17, 29, 32–34, 41, 43–46, 68, 72, 75, 78, 80, 82–95], a “metaphor” of “Web of Causation” [15, 58, 87, 91, 93] was introduced for the description of a complex, interdependent, and multifactorial causation of effects, as stated in [15, 87, 93]. Probably, the term first appeared in the 1960s [15, 93].³⁴

The “Web of Causation” is usually an extensive diagram of the conditions that are necessary for the occurrence (and/or prevention) of a disease in a particular person, as well as in an epidemiological study of the causes of pathology in a population. A web usually looks like rows of rectangles that represent factors and have highly branched and interconnected causal relationships between them (lines) [87, 91, 94]. According to [58], the model is applied to empirical data, including in the WHO.

The web includes direct and indirect causes. Direct causes are “proximal” to pathogenic events (i.e., they are closer to them). Indirect causes are “distal” pathological events (they are more distant, ‘upstream’). For example, obstruction of the coronary artery is a direct cause of myocardial infarction, while the social and environmental factors that lead to hyperlipidemia, obesity, sedentary lifestyle, atherosclerosis, and coronary stenosis are indirect causes. In considering this disease, indirect and direct causes form a hierarchical causal web, which often has reciprocal relationships between factors. The web is divided into levels: the macro-level (social, economic, and cultural determinants), individual level (personal, behavioral, and physiological determinants), and micro-level (organ, tissue, cellular, and molecular causes) [87, 93].

The manuals [87, 94] present an extremely complex scheme of the “Web of Causation” for myocardial infarction, which, in our opinion, can compete in complexity with the scheme of signal transduction of mitogenic and damaging stimuli in a cell (for example, [49]). A no less impressive web is illustrated [91] for the causes of obesity. Persons interested can also find many complex ‘Webs of Causation’ on the English-language Internet.³⁵

It remains unclear how this complex model of the often hardly quantifiable, interpenetrating causal dependences can be used in healthcare practice; unless it is used in a very simplified form or just as an illustration/teaching aid (which is very illustrative, for example, for representing the dubious radiation attribution of cardiovascular and cerebrovascular pathologies after irradiation in small doses [77, 101–103]). The idea arises that there is no adequate verification of the operation of these models for biology and medicine, or even that they are difficult to falsify according to K. Popper’s criterion [11, 52, 64–67].

Probabilistic Cause

This was considered as early as through D. Hume (“probability arising from causes”) [59].

According to the considered M. Parascandola and D.L. Weed, 2001, study [8] on the definitions of causality in epidemiological literature, the “probabilistic” or “statistical” definition of causality is widely used when a cause increases the likelihood of an effect (for example, the development of cancer). The occurrence

of cancer in an individual is partly random (a stochastic, or “nondeterministic” process). In this regard, the probabilistic cause can be neither necessary nor sufficient for the manifestation of the effect, although this definition does not exclude either necessary or sufficient causes. According to [8], a sufficient cause is one that increases the probability that an effect will reach unity, and a necessary cause makes the probability simply greater than zero.

The probabilistic definition of causality is more constructive than the determination of a cause based on a sufficient component. Necessary and sufficient causes can be described in probabilistic terms, however probabilistic causes cannot be expressed from determinate positions. Also important is that the probabilistic definition narrows the number of assumptions about biological mechanisms, since it does not require belief in countless modifiers of a latent effect for each less perfect correlation [8].

It is also possible to use the concept of probability for K. Rothman’s “Causal Pie,” just as it is necessary to keep in mind that components contribute to the probability of an effect and are not determined for it. If one of the components is absent, then the probability of the effect decreases [8]. In 2005, K.J. Rothman and S. Greenland [44] suggested that the presence or absence of one component could be determined via random process; however, in this approach by the mentioned authors, the relationship between cause and effect nevertheless remained determined, although more difficult to predict. Causal relationships often seem probabilistic in practice, however, according to the indicated authors (cited from [72]), this may be due to ignorance of the complete set of components that make up the complex sufficient cause of a pathology.

One of the problems that seems to be against the probabilistic definition is that it cannot explain why, for example, some smokers develop lung cancer, while others do not. That is, “probability” is a kind of euphemism of ignorance. The study [8] resolves this issue, stating that the probabilistic determination also admits the possibility of action of some yet unknown causes, giving the example of a discovered genetic factor that sharply increases the risk of developing lung cancer. Therefore, there is no reason to believe that some cause increases the risk of an effect for each person by the same amount: due to the mathematical continuum of probability, the model allows a wide range of possible effects [8] (see note 32).

Since, as said, it is not possible to distinguish between causal and noncausal associations for the probabilistic determination of causality, a counterfeit element is added to the definition to improve the situation [8].

A Counterfeit Cause

Counterfeit causality is based on the contrast between an effect in the presence of a cause and an effect in its absence. We again find a similar construction in studies by D. Hume: a cause is determined on the basis of whether there would be a second object in the temporal sequence of two objects if there were no previous object [9, 40, 59, 99] (identifies the necessary cause in “the strong sense” [99]).

Such causes can be deterministic and probabilistic [8] (Table 2). For example, according to the definition of N. Cartwright [119] cited in [8], factor *C* causes *E* if probability *E* given by *C* is greater than probability *E* in the absence of *C*, and all other conditions remain constant. As noted above in Table 2, the counterfactual determination of a cause is in the strictest case unprovable, because time does not reverse, and no one can observe the same event, the same individual, or the same population twice with a specific condition (“cause”) and without it [2, 8, 37, 53, 89]. For a strict sense, this problem cannot be solved by experimental approaches involving the use of seemingly completely adequate controls: there are no two absolutely identical cells or organisms, even if they are identical twins. Similarly, a cell and an organism will not be completely identical at different points in time. An absolutely adequate control for a cell and an organism with exposure can be only that cell or the organism without exposure, but in the same time period [89, 91].

Theoretically, even seemingly correct counterfeit “causality” may not be true causality. The S. Greenland and J.M. Robins, 1986, publication [120] provides a speculative example (“bivariant counterfeit” [9]), which, although it seems puzzling, nevertheless is not completely implausible. Let us assume that half of the individuals in a population are sensitive to exposure and can die from it (i.e., they will live only if they are not exposed), and the other half can die precisely because of the absence of exposure (i.e., they will live only if they are exposed). If the exposure is distributed randomly across the population, then the expected average causal effect will be zero: there will be no relationship between exposure and mortality in an infinitely large population. Nevertheless, the observed result for each individual will be due to exposure or nonexposure [9, 120].

All of this looks somewhat abstract and even scholastic from practical biomedical standpoints, but nevertheless counterfeit causality is used quite widely, starting with a well-known experiment of one of the founders of the epidemiological approach, J. Snow, in 1854 in London (when the closure of water taps in specific areas of the city eliminated cases of cholera) [15, 33, 94, 95].

The counterfeit approach was also attempted to be used in radiation and epidemiological studies. For example, a group of authors (J.J. Mangano, J.M. Gould, and others) from the nonprofit organization Radiation and

Public Health Project (New York) published several works that cite data on a decrease in child mortality, including cancer mortality [121] and leukemia mortality [122], not far from American nuclear power plants after shutting down their operation (according to J.J. Mangano et al., during the operation of nuclear power plants, these indicators were increased compared to regions without nuclear power plants [121, 122]). Although the studies of these authors were criticized (“junk” works [123]), were not confirmed in any way by other epidemiological studies (for example, [124]), the very fact of using the counterfactual approach in radiation epidemiology is important³⁶ (another example is a comparison of various indicators in patients that have undergone radiotherapy and, for example, chemotherapy [126]).

Summing up their review study, M. Parascandola and D.L. Weed, 2001 [8] come to the conclusion that *the probabilistic determination of the cause in combination with counterfeit conditions is the most promising for epidemiology*. A similar approach is consistent with deterministic and probabilistic causal models, considering the former as an extreme case of the latter.³⁷ In addition, the mentioned combined model implies fewer assumptions about unobservable natural phenomena, eliminating the need to always establish hidden determinants that make up causes. The determination of causality must allow the possibility that randomness is inherent in some natural processes [8].

Probably, the presented complex causality model has not become universal. Ten years have passed since the publication of the review [8]; a new review by the same author on a similar topic has been published (M. Parascandola, 2011 [72]), and its summary includes the same words: “... epidemiology lacks an explicit, shared theoretical account of causation. Moreover, some epidemiologists exhibit discomfort with the concept of causation, concerned that it creates more confusion than clarity.”³⁸

Nevertheless, the probabilistic causality model is the most acceptable for radiation and epidemiological studies, which are currently concentrated mainly on the stochastic effects of radiation in the low-dose range [21, 22, 24, 25, 50, 51, 77, 101–103, 124]. As for the counterfeit supplement, it is likely to be the lot of disciplines with the possibility of setting up controlled experiments (radiobiology and radiation medicine).

Causality at Different Levels of Biological Organization

According to the authors of [8], there is no reason to argue that causality at one level (for example, molecular) is more significant than causality at other levels, including social. This standpoint does not seem trivial, therefore the priority of causality at different levels of organization has previously been the subject of discussion (see, for example, the work by M. Susser, 1973 [17]). Over the past decades, the question of the

importance of the “black box” strategy for the study of causal relationships in epidemiology has been repeatedly raised [99, 130–132]. On the one hand, critics of this strategy believe that it is necessary to constantly reduce the volume of the “black box,” minimizing the yet unclear mechanisms of causality at the biological level, which are considered as basic. On the other hand, proponents of the primary consideration of social factors for the development of pathologies argue that causal phenomena at the social level cannot be completely reduced to biology or to the determinants of individual behavior, such as, for example, smoking [8].

The two indicated standpoints generally coincide with the two categories of causality (deterministic and probabilistic) [8]. The differences correspond to different types of scientific explanations. So, according to [8], the tendency to give priority to knowledge at the molecular level in developing a causal explanation of an observed association is present among researchers and in the unscientific environment. For example, in the 1950s, an attempt was made to find some specific necessary cause for lung cancer from smoking (such as a specific molecule in cigarette smoke), which would provide only one strict correlation with the occurrence of this pathology [133]. A similar situation arose in the radiobiology of the 1950s–1960s, when researchers of radioprotectors searched for an “atomic bomb pill” in the figurative expression, which also wandered later into respective circles. Among biologists, this reductionism has been criticized; moreover, it has been assumed that causal conclusions in epidemiology should integrate all multilevel events varying from molecular to social [8, 134].

SCIENTIFIC, PRACTICAL, AND SOCIAL OBJECTIVES OF EPIDEMIOLOGY AND STUDIES OF CAUSALITY

Some authors point to the difference between the scientific or “logical” definition of causality and its more flexible practical definition, since it is believed that the goals of science and healthcare are different [8, 39]. The main goal of healthcare is “to intervene to reduce morbidity and mortality from pathology” [118] (cited from [8]). The sociological area of epidemiology was developed mainly by M. Susser [134]. As is usually believed, the main goal of science is to explain the world [1, 2, 8] (“reduction of uncertainty” [2]). In opposition to M. Susser, K. Rothman et al. believed that epidemiologists should strive for scientific objectivity and not allow themselves to be guided by the specific goals of healthcare [135].

Therefore, scientific researches, especially fundamental ones, may or may not serve as the basis for effective health protection strategies [8].

Based on the latter fact, the recommendation was developed that epidemiology should abandon the traditional scientific concept of necessary and sufficient

causes in favor of a broader concept that is not related to strict determinism and has a higher practical significance [8, 11] (such as the concept of risk; see note 27). Although probabilistic causality models are not completely accurate, they can provide a faster description of phenomena, which is important in terms of benefits [136] (cited from [8]).

At this point, it is appropriate to briefly consider the question of what can be considered scientific from a public and social point of view. Sociological researches in the field of science have become the basis for developing the concept of “Sociology of Knowledge” (SoK), according to which science is also social activity [2]. The question of what research objects in the world cannot be resolved without taking into account how a social group consisting of scientific researchers (“scientists”) imagines these objects [2]. Entities, whether these be electrons or DNA, cannot be attributed a role in our world, regardless of symbols and attached meaning, and the specifics of symbolism depend, in particular, on the social status of the one who has developed them [2, 9]. Therefore, while the traditional philosophy of science had special criteria for distinguishing between the scientific and unscientific, such as K. Popper’s criteria of falsifiability [2, 11, 52, 64–67], the “Sociology of Knowledge” applies social criteria for these purposes [2].

As noted by a famous philosopher of science Thomas Samuel Kuhn (T.S. Kuhn; 1922–1996) [137], when researchers must choose between competing theories, two people can come to different conclusions, although they may profess the same standard evaluation criteria. This depends on the inclinations, personality, biography, and other characteristics of the individual. The following examples can be added: a psychometric study (2001) by Australian and New Zealand epidemiologists on which causality determination methods are prioritized [138], as well as a survey (2009) of specialists from different countries regarding the preference of a particular “dose-effect” model for radiation safety (linear nonthreshold, superlinear or sublinear threshold safety) [139].

To eliminate reliance only on the subjective opinions of authorities and experts, “evidence-based medicine” was developed in the 1990s [55].

A foreign academic manual on the philosophy of science published in 2007 [2]³⁹ says the following: “There is no question about what the theory represents (either nature or culture), but rather it is a question of negotiation between different scientific groups with regard to what will be considered to compromise facts. Hence, ... the issues are not the relationship between theory and nature/culture (epistemological and representational), but what scientists regard and treat as real (ontological and processual).”⁴⁰

Similar positions can be partly found in the documents of the UNSCEAR, for example, in the

UNSCEAR-2012 published in 2015 (highlighted by the author below) [21]:

“The scientific method does not operate in isolation, but is conducted by the scientific community, which has specific internal norms to guide the activities of scientists in applying the scientific method. These norms include truthfulness, consistency, coherence, testability, reproducibility, validity, reliability, openness, impartiality and transparency [140]... Incorporation of nonscientific concerns ...—may or may not use conditional predictions. In this case, account is taken of *norms external to science such as social responsibility, ethics, utility, prudence, precaution and practicality of application* ... The precautionary approach⁴¹ can be described in the following way: “When human activities may lead to morally unacceptable harm that is scientifically plausible but uncertain, actions shall be taken to avoid or diminish that harm”⁴² [21].

In our opinion, this principle may have disadvantages of a significant conjuncture of unscientific approaches in the formation of practical conclusions on an allegedly scientific basis, but beyond the scope of scientific ethics. So, criteria are not always available for an accurate assessment of the social significance of a problem under study, and they clearly will not be, so to speak, understated by the authors.

In the widely acclaimed lecture by Michael Crichton, which was later published in *The Wall Street Journal* [151] (see also an in-depth discussion of this issue in [55]), the author considered consensus science “as an extremely pernicious development that ought to be stopped.” He emphasized, “Let’s be clear: the work of science has nothing whatsoever to do with consensus. Consensus is the business of politics. Science, on the contrary, requires only one investigator who happens to be right, which means that he or she has results that are verifiable by reference to the real world... There is no such thing as consensus science. If it’s consensus, it isn’t science. If it’s science, it isn’t consensus.”⁴³

CONFOUND FACTORS AND FALSE ASSOCIATIONS

“Confound factors,⁴⁴ i.e., “confounding” factors in the main translation from English [15, 32, 33, 153] (sometimes there are “confounding” [94]), with already widespread slang calque “confounders” or “confoundings” [33, 153, etc.], represent one of the deviations from true causality, which can lead to systematic subjective deviations or biases [154] (sometimes the term “systematic error, confounding bias” is met) [15, 155]).

According to the definition from the fundamental dictionary of epidemiology [15], a confounder is “a variable that can cause or prevent a studied outcome, but is not intermediate in the causal chain and is associated with a studied effect.” The manual [153]

addresses “strengthening, weakening, perversion” instead of “cause or prevent.”

At this point, we will not dwell on the theory of confound factors, which is fairly common material in manuals on epidemiology [3–5, 15, 29, 50, 51, 83, 87, 89, 91, 94, 95, 153] (see also reviews on definitions, theories, and history of the formation of the concept together with a retrospective look at relevant studies [9, 120, 152, 154]). Likewise, we will not dwell on the methodological and statistical approaches that can eliminate the effect of potential confounders (randomization, stratification, standardization, etc. [33, 120, 153]).

We give only a summary of the examples of the influence of confound factors leading to statistically significant but false associations from various fields of natural sciences and sociological disciplines (Table 4—nonradiation factors; Table 5—radiation exposure).⁴⁵ Although the points reflected in Table 4 suggest some sensationalism, all the relevant links are nevertheless strictly scientific (on the Internet, we can find many examples of graphs with false associations, however almost none are provided with an indication of sources).

Data summaries with similar false associations can be made, not only for confound factors, but also for the most varied types of subjective biases. These (more than a dozen types can be found) are also considered in all epidemiology manuals and have been identified in many relevant studies [2, 15, 18, 20–22, 24, 25, 29, 32, 33, 43, 44, 46, 50, 51, 89, 91, 94, 95, 101–103, 106, 153, 172, 173]; we will not dwell on them.

Many biologically unrelated associations can occur when searching for genetic markers.⁴⁶ The A.V. Rubanovich and N.N. Khromov-Borisov, 2013 [195], study concluded that the predictive and classification efficiency of the results of most published associative genetic studies was low.

The mass of false associations listed in Tables 4 and 5, which often includes even statistically significant trends for dependences on radiation dose (!), unequivocally indicates the need to use certain epidemiological criteria or at least generally accepted rules of assessment that would make it possible to determine the degree of truthfulness of the identified relationships. The criteria for determining the causation of effects from exposures as well as the relative contribution of epidemiology and experimental disciplines towards them are planned for presentation in the following reports.

CONCLUSIONS

The presented Report 1 has considered, *first*, the conceptual formulation of the problem and the purpose of the report cycle, as well as preliminary philosophical and scientific concepts that are necessary for the subsequent understanding of the main problem—

Table 4. Examples of false associations caused by confound factors, for nonradiation impacts

Statistically significant association or dependence	Explanation (confounder)	References
Ice cream consumption—frequency of deaths from drowning	Climatic period	[26, 27]
Ice cream consumption—frequency of deaths from homicides	Climatic period and crowding in recreational areas	[28]
Imports of apples—number of divorces in the United Kingdom	No explanation; randomness is possible	[156]
Growth in bio-food sales*—increase in the frequency of autism in the United States	No explanation; randomness is possible	[91]
Living in the coastal resort towns of England— increase in the mortality rate compared to the general population	Resort towns attract retired and elderly people (the older population)	[157]
Compared to Caucasians, African Americans are more intense consumers of the “craig” hard drug	This is attributed to greater poverty of African Americans	[157]
Male sex—frequency of lung cancer	Smoking	[5, 29]
Smoking—frequency of accidents	Alcohol	[35]
Smoking—frequency of suicides	Probably a way of life. The likelihood of association was identified in the studies as “negative control” (it was found that there was a similar dependence with the frequency of deaths from homicides, which deliberately cannot depend on smoking) [18]	[158]
Smoking—frequency of cirrhosis	Alcohol	[35]
Yellow-brown fingers—frequency of lung cancer	Smoking	[9]
Occupational exposure—frequency of various pathologies	Smoking	[159]
Alcohol consumption—frequency of lung cancer	Smoking	[94]
Coffee consumption—frequency of lung cancer	Smoking	[160]
Coffee consumption—frequency of pancreatic cancer	Smoking	[161]
Coffee consumption—frequency of infarcts	Education, income, profession, etc., that determine lifestyle**	[33]
Coffee consumption by pregnant women— frequency of miscarriages or newborns with somatic defects	The magnitude of the effects of hormonal changes during pregnancy (nausea, vomiting, and appetite disorders) is <i>inversely</i> proportional to the likelihood of a miscarriage or a newborn with defects. The more pronounced are the effects of hormonal changes, the less coffee pregnant women drink (and vice versa: in a relatively low concentration of pregnancy hormones, they drink more coffee)	[33]
Success of a false operation as a control in clinical trials	In the formation of groups: lack of homogeneity, consideration for the sex-age distribution, false double blindness***, lack of postoperative double blindness, understatement of psychological and psychiatric factors	[162]
Order of child birth in the offspring—frequency of birth of newborns with Down’s Syndrome	The age of mother	[89, 163]
Social background—frequency of newborns with Down’s Syndrome	The age of mother during pregnancy can be determined by social status	[44]

Table 4. (Contd.)

Statistically significant association or dependence	Explanation (confounder)	References
Intensity of hormone replacement therapy—decrease in the frequency of deaths from cardiovascular diseases in women	Probably a way of life. The likelihood of the association was identified in the studies as “negative control” (a similar dependence was revealed—decrease in the frequency of deaths from accidents, homicides and suicides, which is absurd in terms of causality) [18]	[164] (in addition, see discussion in [33, 37])
Pregnant women taking drugs—risk of defects in newborns	Age, fertility, smoking, alcohol, body mass index, reproductive failure, previous pregnancies, previous births of children with defects, socioeconomic, racial, ethical, geographical and seasonal factors. In rare cases, gender of the newborn and multiple pregnancies	[165]
The correctness of the transcriptome analysis for confounders in the study of effects	Individual characteristics of the object, composition of the sample, and time of day of determination	[166]

* That is, without pesticides, synthetic additives, etc.

** We will add the smoking factor.

*** The condition for testing a medical intervention is the “blinding” of participants (blinding, Eng.), i.e., hiding from them to which group in the study they belong. The term has its origins in the work of A. Lavoisier and B. Franklin in the late 18th century during experimental testing of therapy through hypnosis (the experimenters then blindfolded the participants so they did not see what was happening). If the treatment is known to the researcher, but not to the participant, then this is a “blind trial.” If the doctors working with the patients are also ignorant, then this is a “double-blind trial.” Finally, “triple-blind trials” are those in which the randomization result is hidden not only from the patient and the researcher but also from the statistician that analyzes the data [33].

the meaning, essence, and possibility of practical application of the criteria (rules, principles) for establishing the truth of associations detected in biomedical disciplines. The idea and relevance of our study was determined, first of all, by the weak penetration of specific methodologies for proving causality into experimental and descriptive disciplines related to the influence of the radiation factor (especially in Russia).

The formulation of a comprehensive definition of causes and causality is not a very simple task, despite its apparent clarity. Five types of definitions varying from the simplest explanatory definition to complex integrated scientific definitions for deterministic and stochastic effects have been identified. These five types of concepts, many of which go back to the positions of famous philosophers (mainly David Hume [59] and others, in particular John Stuart Mill [62]), were fragmented in epidemiology as early as the early 2000s [8]. However, judging by the publications of researchers specialized in this problem [5, 29, 44, 72, 84], they have not become a universal base platform for studying biomedical effects from various agents and impacts even 10 or more years after. There is criticism of certain types of definitions, which is related to restrictions for use in specific areas of natural sciences.

Second, this report has presented a set of provisions from significant sources (publications of researchers of the theory of causality, manuals on the philosophy of natural sciences, and UNSCEAR documents) that reveal the scientific, practical, and social goals of epi-

demiology and, in fact, any scientific researches that are important for human life and activity. These goals are primarily associated with evidence of the truth of the revealed dependences of effects from agents and impacts, however methods for their achievement can be based on different rules and ethical foundations proceeding from the existing fundamental scientific or social tasks. In the second case, the “precautionary principle” is used, and research standards developed for scientific application are significantly narrowed and simplified, being largely replaced by one thing: the prevention or, at least, reduction of risks, even if the reality of the latter does not have rigorous scientific evidence.

This fragment of Report 1 is necessary for ascertaining which areas researcher work may be of primary importance in proving the reality of effects—whether it is important only for scientific areas, or also for public and social. This article will help find out what limits of scientific evidence should be reached depending on one or another socially significant conjuncture.

Finally, *third*, this publication yields examples of false, but statistically significant, associations from a wide variety of biomedical as well as public and social spheres. The falsity was caused by third—confound—factors. Some of the information presented looks sensational, and some is absurd. There is also one that is perceived as very plausible (for example, the relationship between smoking and suicides, coffee consumption and miscarriages, hormone replacement therapy

Table 5. Examples of false associations caused by confound factors in radiation exposure

Effect	Explanation (confounder)	References
The lower rate of mortality from various pathologies for workers in the nuclear industry (and other harmful industries) in comparison with the corresponding gender and age group of the general population	“The healthy worker effect” (hiring relatively healthy individuals for work in harmful industries)	[15, 33, 167–169]
Association between irradiation dose and lung cancer frequency in nuclear workers in the United States	Imitation of “dose–effect” relationship with smoking	[170]
Association between irradiation dose and frequency of smoking in women affected by the atomic bombing in Hiroshima (not in men and not in both sexes in Nagasaki)	Imitation of “dose–effect” relationship with smoking	[171, 172]
Association between irradiation dose and rate of mortality from nonradiation-related pathologies (cirrhosis of the liver, cancer of the oral cavity and pharynx, psychoses and external causes) among nuclear workers in France	Imitation of “dose–effect” relationship with alcohol	[169]
Association between frequency of cancers and leukoses in children and dose received by their fathers employed at the nuclear center in Sellafield	Imitation of the “dose–effect” relationship by mixing of the population with newcomers in the early period (viral infections, etc.)	[173]
Increased rate of mortality from lung cancer due to radiation exposures (meta-analysis: nuclear industry, medical exposure, etc.)	Below a dose of 1 Gy, the effect was not found in “never-smoked” individuals	[174]
Mortality from solid cancers for Japanese nuclear workers with adjustment for confound factors	Excessive relative risk (ERR) compared to background frequency is 0.78. Adjustment for smoking is 0.31; adjustment for the beginning of professional employment (education) is 0.35; smoking + education – minus 0.09	[175]
Deviations and pathologies in children whom descended from the liquidators of the Chernobyl accident	Effects of nonradiation factors in fathers (heavy metals, gasoline, fuels, and lubricants; solvents and other organic compounds; medications; drugs; alcohol; smoking). Psychogenic stresses in fathers and mothers during pregnancy and later in families [175, 176]. The possibility of stress from an unwanted pregnancy that leads to disorders and pathologies in the offspring [177, 178] (etc., see [175]*)	[176, 177]
Effects of the Chernobyl nuclear power plant accident	Changes in the socio-political situation after the disintegration of the Soviet Union; the main contribution is social and economic change and significant stress**	[25]
Effects of the accident at the “Fukushima-1” NPP	Chemical contaminants resulting from the earthquake and tsunami can be serious confound factors for the carcinogenic effect of radiation	[180, 181]
A 9.5-fold increase in the incidence of congenital malformations compared with the control level in the Lyudinovsky district of the Kaluga oblast after the Chernobyl nuclear power plant accident	Lack of data on the background level specifically in this region, which differs from other regions of the Kaluga oblast in a more developed industry and low level of employment in agriculture (industrial factors)	[182]

Table 5. (Contd.)

Effect	Explanation (confounder)	References
Frequency of thyroid cancer after radiation incidents, mass medical exposures, or simply mass examinations of this organ, which affect large populations	Screening effect. Sevenfold increase in the examination of persons whom received radiation of the head and neck (the United States) [183]. An increase of 1.4 and 2.6 times (for women and men) among the Swiss population in the 1990s compared to the 1970s [184] (there is much similar data for other countries, for example, the United States [185]). In addition, a 50-fold increase after the accident at the Fukushima-1 nuclear power plant in central Japan [181, 186]. For children of the Chernobyl contingent—2.5 times [187] and 1.4–1.9 times [188] at the expense of screening (and other sources)	[22, 24, 25, 171, 183–188]
Increase in the frequency of thyroid cancer after therapeutic impacts with ¹³¹ I on account of noncancerous thyroid pathologies	Without radiation – a multiple (by orders of magnitude) increase in the risk of development of thyroid cancer in hyperthyroidism, Graves' disease, benign nodular neoplasms in the organ, etc. (“reverse causation”; see above)	[23, 189]***
Increased incidence of cancers and leucoses after CT in child age	The factor of “reverse causation”: CT was more often performed in those who were suspected to have malignant neoplasms	[19–22]
Induction of secondary leucoses after radiotherapy and chemotherapy of bone marrow in children	The statistically significant trend of increase in secondary leucoses depending on irradiation becomes insignificant after making adjustment for chemotherapy	[190]
Induction of breast cancer after breast radiotherapy on account of other pathologies	Decrease in the latent period depending on the age at the time of exposure (26 years for exposure before the age of 30 years and 22 years for exposure at the age over 30 years) may be a confound factor in the estimate of age-related radiation dose	[191]
Genome instability after irradiation of certain cells in small and close doses	Nonradiation-induced aging of cells during subsequent passages rather than the radiation factor	[192–194]

* Our earlier reviews [176, 177] also mentioned a Russian publication (online conference materials) with information on the spectrum of psychosomatic pathologies in children born from unwanted pregnancies (K.Sh. Brekhman and G.I. Brekhman, 1998). The source has been cited more than once on the Internet, however the original on <http://www.psymama.ru> is no longer available.

** “... the severe disruption caused by the accident, confounded with the remarkable political changes that took place in the Soviet Union and the new republics, resulted in major social and economic impact, and great distress for the affected populations” [25].

*** Our summary on the topic includes many sources related to the period from 1985. Here, only two references to reviews are presented.

and a decrease in the risk of cardiovascular pathologies, the dependence of the frequency of malignant neoplasms in nuclear workers and their offspring on irradiation dose, etc.; Tables 4 and 5). Nevertheless, all turned out to be false, although they do not seem to be “worse” by external signs than many true associations published in scientific sources.

The given examples of the effects of confound factors that form false associations indicate the need to use standard criteria for assessing the truth of causal relationships. Several complexes of these criteria are known, one for infectious pathologies (initially the

Henle–Koch criteria with further modern modifications by other authors, up to the universal Evans complex), and the other two for noninfectious pathologies (the Hill and Susser criteria). This material is planned for consideration in subsequent reports.

Conflict of Interests and Possibility of Subjective Biases

There is no conflict of interest. This study that was conducted within the broader budget topic of research development at Russia's Federal Medical Biological Agency was not supported by any other sources of funding. There were no timeframes, official require-

ments, restrictions, or other external objective or subjective interfering factors in carrying out the study.

NOTES

(1) We can add to this the words of King Lear: “Nothing will come of nothing” The source of these phrases is the treatise of Titus Lucretius Cara “On the Nature of Things”, which reproduces the teachings of Epicurus.

(2) “... most researchers would find it difficult to define the words in anything but a circular fashion; causes are conditions and events that produce effects, and effects are conditions and events produced by causes” [9] (hereinafter, the translations were presented by the author—A.K.).

(3) “I have no wish, nor the skill, to embark upon a philosophical discussion of the meaning of “causation”” [13].

(4) M.W. Susser, who, in fact, made probably one of the largest contributions to the rules for establishing causality [9, 14] (for more detail, see Report 3), is not mentioned in any Russian-language manuals on epidemiology and evidence-based medicine known to us. The exception is the *Epidemiological Dictionary* [15], which is a translation of the famous Oxford publication edited by J.M. Last (2001). The domestic publication by V.V. Shkarin and O.V. Kovalishena, 2013 [16] was found to mention the surname of M. Susser in the context of “modifications of Hill’s criteria of causality” (which is incorrect) with reference only to the indicated translation of the dictionary [15].

(5) “... any factor, whether event, characteristic, or other definable entity, so long as it brings about change for better or worse in a health condition” [17] (cited from [9]).

(6) “... something that makes a difference” [11].

(7) In the majority of relevant studies and documents, the essence of the term “reverse causation” is assumed to be clear and is not explained in any way, just specific example(s) is given. The single definitions of “reverse causation” that are known to us are not clear and understandable, and, therefore they are probably also always accompanied by a real example. So, in [18], the phenomenon has the following explanation: “Pre-existing symptoms of the outcome that influence the exposure could generate the observed associations.” A well-known example is presented that the curve of alcohol-related mortality is J-shaped, i.e., complete nondrinkers live moderately less than drinkers (it is understood that nondrinkers may not drink because of their initially worse health status).

(8) The assumption that the male sex determines the occurrence of lung cancer, which seemed anecdotal from the standpoint of modern knowledge, did not seem as such relatively recently. For example, in 1947, Dr. Evarts Graham in Missouri conducted therapy for this pathology by administering female sex

hormones to patients, since it was assumed that women were protected from lung cancer (the therapy was unsuccessful) [30, 31].

(9) There are many reviews on the history of establishing a causal relationship between smoking and lung cancer. Here, we present only a single source—the work of one of the main researchers of this problem in the 1950s [35]. A more detailed review is planned in Report 2.

(10) Information that the main criteria (rules) of causality for noninfectious pathologies were proposed by several authors as early as the 1950s is not presented in most known English-language sources of the topic, including monographs, reviews, study guides (in particular, online), and fundamental dictionaries on epidemiology (for example, [3, 15]), not to mention Russian-language publications. Information of this type is found only in studies specifically devoted to the historical aspects of the formation of causality criteria. There are also few of them. We can mention the relatively recent reviews of the main researchers: A.S. Evans, 1976 [34]; D.L. Weed, 1988 [36]; and M. Susser, 1991 [11] (the original monograph of the last author as of 1973 [17] is not available to us). Some information with reference to the study by M. Susser (1991) [11] is presented in one of the latest official reports of the US Department of Health on the health effects of smoking (2004) [37]. This information is also available in relatively few reviews and manuals of later authors that discussed the philosophical and epidemiological problems of causality in the 2000s–2010s [38–41]. We will talk about the deplorable state of priorities in this area further, especially in Report 2. The “criteria of (Bradford) Hill” formally are not Hill’s criteria. Besides us, the question of inexplicable noncompliance with copyright ethics by the most famous founders of the causation theory was also raised in the H. Blackburn and D. Labarthe, 2012, historical study [42].

(11) Again, without reference to sources, i.e., to the authors who proposed those principles. This can be seen from the document [43] and is also emphasized in studies by M. Susser, 1991 [11] and H. Blackburn, D. Labarthe, 2012 [42].

(12) “... these guidelines have generated a talmudic literature on their nature, logic, and application” [9].

(13) This “evidence” is often limited when advertising nutritional supplements, etc.

(14) At this point, we can recall the situation in the field of aging genetics. The authors receive Nobel Prizes for extending the life of the nematode and fruit fly, while the discovered genes that may be responsible for the life expectancy are rarely significant when studying the genomes of “long livers,” if they are significant at all (at least, this was the state of affairs in 2012 [48, 49]). Of course, extending the life of the nematode and fruit fly could also be regarded as helping nature, however, in our opinion, the awarding of the Nobel Prizes for this is still somewhat premature.

(15) In 2016 and 2017, the author of this publication made two almost similar reports at the FSBI SSC Burnazyan Federal Medical Biophysical Center of Russia's Federal Medical Biological Agency (the Academic Council) and at the FSBIS "Federal Research Center of Food and Biotechnology" (the Council of Young Scientists and Specialists). The topic was the criteria of causality and the hierarchy of biomedical disciplines in giving evidence of effects from impacts. In the second organization, only one representative of a fairly young audience knew about the existence of causality criteria. In the first, in short, the situation also left much to be desired.

(16) "... deductive logic could never be predictive without the fruits of inductive inference" [57].

(17) "... argued that 'true causes' are both necessary and sufficient to produce a given effect. It further followed from Galileo's analysis that true causes are universal" [58].

(18) "... reasoning was to imply the principle of causality from the assumption that it is among the conditions of every experience" [5].

(19) "Rules by which to judge of causes and effects" [38, 59].

(20) Mill's *Eliminative Methods of Induction* (System of Logic, 1843) [63].

(21) "... alluding to Hume, stressed the psychological nature of these concepts [causality] and pointed out that "in nature there is no cause and no effect" and that these concepts are results of an economical processing of perceptions by the human mind" [5].

(22) "... that causation 'is a relic from a bygone age, surviving, like the monarchy, only because it is erroneously supposed to do no harm'" (Russell, 1959, p. 180) [2].

(23) "... refuted Hume's subjective view of the world by the demonstration, which he attributed to Immanuel Kant, that a priori knowledge exists independent of experience. Russell himself showed that relationships, too, can exist independent of experience" [11].

(24) We note that we have managed to find a consistent and complete summary of the development of the concept of causality over the centuries in almost no manuals on the philosophy of science (English-speaking manuals and many voluminous Russian-speaking manuals, for example, those cited above [1, 2, 52, 54, 56, 60, 63] and others [69, 70]). The presented information is discrete and fragmentary, often mixed; personalities are presented in different sources in a dotted manner (sometimes they are included; sometimes they are not); a complete summary was not found anywhere (this caused an excessive number of general references in Table 1, since the views of all philosophers involved in the problem of causality were not considered in any manual). As mentioned, the exception known to us is only a very brief chapter in the English monograph of 2002 [68], however it is of

philosophical rather than natural scientific nature and does not contain provisions regarding modern philosophers, ending with the views of J. Mill.

(25) Perhaps, due to V.I. Lenin's [71] stereotypes, even the mention of D. Hume is missing in some voluminous Russian manuals on the philosophy of science (for example: [69]—2006, 136 pages; [70]—2007, 731 pages).

(26) As indicated, Table 2 presents the data from the review [8] (additions from [5, 29]) with a fairly in-depth search for relevant sources in PubMed. Of course, the definitions of causality in biomedical disciplines have been the subject of consideration in all epochs (we recall once again the reference to the question of "causation" of fever in connection with the arrival of a frigate in Liverpool in 1861 [12]). This problem was not left out of account in domestic medicine either; it is enough to recall the monograph *The Problem of Causality in Medicine (Etiology)* by I.V. Davydovskii, which was published in 1962 [74]. Its author noted as early as this time that "a huge philosophical and natural-historical literature that covers the problems of causality in biology has been accumulated"; moreover, the cited sources (which were, however, few) included the whole collection *The Problem of Causality in Modern Biology* (Moscow: Izd. SSSR, 1961). Thus, mass profile discussions took place long before the "talmudic period" [9] after 1965 [13]. For our time, one cannot but mention a very comprehensive section on the causality of effects in the manual on epidemiology of the leading Russian expert in the field of evidence-based medicine V.V. Vlasov (2006) [33].

(27) "... risk factor epidemiology cannot be considered as a scientific discipline because it aims at concrete usefulness rather than abstract truthfulness" [80]. The fact is that a risk factor is not necessarily a cause (and science studies precisely specific causes, "reducing uncertainty" [1, 2, 8]). When noninfectious pathologies with long latent periods are considered, the concept of "risk factor" is used in medicine to determine the intervention strategy. But the categories of "risk factor" and "cause" do not coincide. For example, elevated cholesterol is a risk factor, but not the cause of coronary heart disease; childhood obesity is a risk factor, but not the cause of diabetes; thickening in the breast tissue is a risk factor of cancer of this organ, but not always the cause. The indicated difference in concepts is critically important because it can be difficult to obtain evidence of a statistically significant relationship between *a risk factor and real effect* [81].

(28) In total, at the beginning of 2018, PubMed included 382 publications by Professor of Epidemiology and Medicine at Boston University Kenneth J. Rothman (born in 1945), most of which are devoted to the epidemiological effects of a wide variety of impacts. Starting with the work of 1969 and until very recently (2017), this author has also been studying

urgent aspects of the theory of epidemiological researches [85].

(29) There are many sources on epidemiology with illustrations of Rothman's causality model on the foreign Internet. In RuNet, they are few (perhaps we have mentioned main ones [95, 96]).

(30) The following intricate example of INUS is given in [2]. Imagine that a person drinks a lethal dose of poison without taking an antidote, and, when his stomach does not react by vomiting, the person dies. What is the cause of death? Did the person die as a result of having ingested poison, from not taking an antidote, or from his stomach not having reacted by vomiting? The mere ingestion of poison is not enough: many people took the poison without that lethal outcome (because they vomited). But poisoning is part of a combination of conditions that together are *sufficient* for the lethal outcome. In addition, proceeding precisely from this complex of lethal conditions (without taking into account other complexes that are sufficient for death from other causes), poisoning alone is not superfluous: deaths from poison do not occur under circumstances when the poison was not taken. Thus, the intake of poison is an insufficient, but not superfluous part, of an unnecessary, but sufficient condition for death.

(31) The "conventionality" of our modification consists in the different number of sectors in the circles-"pies" compared to Rothman's original (1976–2005 [10, 44, 82]), which is not important for essence. Probably, in connection with copyright even for a simple scheme, some Western authors, introducing this as an illustration in their reviews and manuals, took the path of modification, having replaced the original three circles with five sectors in each (from [10, 44, 82]) with other numbers of circles and/or sectors [58, 87]. Admittedly, a number of sources [89, 92–95] reproduce the scheme authentically to the original.

(32) Some authors believed that the epidemiology of noninfectious pathologies should proceed more from the concept of risk, chance, than causality (due to the uncertainty of the probabilistic definition) [72]. The concept of risk as a continuously changing category does not correspond to this term for a particular person. For an individual, the risk of an effect (disease) has only two meanings: zero and unity ("fatalism prevails," as stated in [8, 84]). The use of intermediate values for an individual risk parameter is only a method of estimating an indicator for an individual based on the average risk for many other, presumably, similar individuals. The actual risk for an individual depends on whether a sufficient cause has been formed for him or will be formed, while the average risk for a group indicates only the proportion of persons with *already* formed sufficient causes. Individual risk can be considered as a statement about the likelihood of the formation of a sufficient cause of a disease in the corresponding time [10]. Nevertheless, as stated

[8], all events are completely determined at an individual level, because their appearance or lack of appearance is completely determined by existing circumstances. The same source [8] notes that many epidemiologists profess strict determinism, and that this underlies the widely cited work of K. Rothman (1976) remains proved [10].

(33) To give an epidemiological example that falls under the first approach, [82] presents the data of two American case-control studies on the risk of stroke in young women after taking oral contraceptives. In the first study, a relative risk of 2.3 and 8.8 was obtained for different types of stroke, and, in the second, there was a risk of 26 for all types. However, there was a difference in the sample: the first work performed for patients from 91 US clinics reflected the general population as a whole, and the second work reflected a smaller group, from which individuals with hypertension, diabetes, and rheumatic cardiac pathologies were initially excluded. That is, in the second case, a healthier group with an initially lower level of strokes was studied. Similar aspects can be important for radiation epidemiology, for example, when studying cardiovascular and cerebrovascular pathologies in workers in nuclear and other harmful industries [77, 101–103]. In these cases, it is prohibited not only to compare the frequency of the parameter with the indicator for the general population (due to the "healthy worker effect" [15, 33]), but also to compare the magnitude of the increase in the frequency of any pathology, for example, after exposure of a cohort of workers and general population. This is so since in equal exposures the relative frequency of pathology in workers can increase more than in the corresponding gender and age group of the population, because the background level of the indicator in the workers is lower. As a result, it would be that, despite the "healthy worker effect," the sensitivity of these workers to pathological effects would seem higher.

(34) In [58] the development of the concept of the "Web of Causation" is associated with the name of Christopher J.L. Murray; however, we have found the works of this author on the topic only for the 2000s. The American Epidemiology Dictionary [15] attributes the first use of the methodology to T.R. Dawber et al. in 1959 (coronary heart diseases) [116]. However, the original [116] contains neither webs or illustrations nor the corresponding term, but simply tables of multifactorial causation of pathologies. Finally, a historical review on the topic [93] does not name the mentioned founders, but presents a number of other researchers, and primacy is given, of course, to the monograph published in 1960 by Brian MacMahon et al. [117]. A review of one of the leading analysts of causality in epidemiology [72] also attributes the introduction of the term "Web of Causation" to the monograph by B. MacMahon et al., but to the one which was published in 1970 [118] rather than in 1960 [117] (the text of [72] also cites [117]). The mono-

graphs [117, 118] are not available to us; judging by the illustrative data included in [93], priority nevertheless must be given to the publication in 1960.

(35) The pinnacle of “imagery” is probably the Internet scheme “Web of Causation” in the form of a spider web with a spider disease that lies in the middle of the structure in wait for a victim of excesses, poor environment, and poor heredity.

(36) The studies by J.J. Mangano and J.M. Gould (the founder of the “Radiation and Public Health Project”) et al. are not cited in documents of the main international organizations (UNSCEAR, ICRP, BEIR, COMARE, NCRP); we did not find their consideration in any radiation-epidemiological scientific literature. The exception was the publication of A.V. Yablokov “The Myth of the Safety of Small Radiation Doses” in 2002 [125]. Although the surname of J.J. Mangano is presented in the text of this brochure with an error, nevertheless, the source of A.V. Yablokov was the only one in which we managed to find an attempt to use an exact counterfeit approach to causality in radiation and epidemiological studies.

(37) During a certain intensity of some exposures and established circumstances, stochastic causes can be almost deterministic. So, in syndromes caused by DNA repair defects, the frequency of malignant neoplasms can be very high: for example, 40% by the age of 20 years in case of the Nijmegen breakage syndrome [127]. Another striking example is chronic exposure to asbestos. A case was described (the United States) [128], which took place in the family of a worker employed in the production of asbestos insulation for pipes, who came home for many years in work clothes that his wife shook and washed. The material of the cotton bags in which asbestos was transported was used in the family after washing laundry such as diapers for the son. The mother died of mesothelioma (a main malignant tumor of the lung due to asbestos exposure [100]) at the age of 49 years, and the son died at the age of 32 years. Previously, the uncle of the mother who lived in this family (he also worked in asbestos production for some time) fell ill with mesothelioma at the age of 43 years. The father died at the age of 53 years from asbestosis with concomitant cirrhosis. The younger daughter seemed to be unaffected by pulmonary neoplasms. Thus, the frequency of the specific tumor in the family was 60%. In another similar study, three out of five family members of an asbestos production worker also developed lung cancer or mesothelioma (out of three daughters, the two youngest lived) [129].

(38) “... epidemiology lacks an explicit, shared theoretical account of causation. Moreover, some epidemiologists exhibit discomfort with the concept of causation itself, concerned that it creates more confusion than clarity” [72].

(39) There is a reissue of the manual [2] published in 2015.

(40) “There is no question about what the theory represents (either nature or culture), but rather it is a question of negotiation between different scientific groups with regard to what will be considered to compromise facts. Hence, ... the issues are not the relationship between theory and nature/culture (epistemological and representational), but what scientists regard and treat as real (ontological and processual)” [2].

(41) At the UNSCEAR-2012 [19], the “precautionary principle” was given a number of references of varying weight, including two notes in the *New Scientist*, as well as documents of various organizations [141–143] and other significant sources [2, 144–148] (1990–2008). We can add very recent publications on the topic, for example [149, 150].

(42) “The scientific method does not operate in isolation, but is conducted by the scientific community, which has specific internal norms to guide the activities of scientists in applying the scientific method. These norms include truthfulness, consistency, coherence, testability, reproducibility, validity, reliability, openness, impartiality and transparency [140]... Incorporation of non-scientific concerns ...— may or may not use conditional predictions. In this case, account is taken of norms external to science such as social responsibility, ethics, utility, prudence, precaution and practicality of application ... The precautionary approach can be described in the following way: “When human activities may lead to morally unacceptable harm that is scientifically plausible but uncertain, actions shall be taken to avoid or diminish that harm” [21].

(43) “I want to pause here and talk about this notion of consensus, and the rise of what has been called consensus science. I regard consensus science as an extremely pernicious development that ought to be stopped cold in its tracks. Historically, the claim of consensus has been the first refuge of scoundrels; it is a way to avoid debate by claiming that the matter is already settled. ... Let’s be clear: the work of science has nothing whatsoever to do with consensus. Consensus is the business of politics. Science, on the contrary, requires only one investigator who happens to be right, which means that he or she has results that are verifiable by reference to the real world. In science consensus is irrelevant. What is relevant is reproducible results. The greatest scientists in history are great precisely because they broke with the consensus. There is no such thing as consensus science. If it’s consensus, it isn’t science. If it’s science, it isn’t consensus!” [55, 151].

(44) From the medieval Latin word “confundere”—“to mix together” [152, 153].

(45) The most common compound factors are gender, age and place of residence [25, 33, 94, 152, 153].

(46) The currently popular trend. Sensational data about allegedly finding “genes of religiosity”, “genes of intelligence”, etc., are often published in the media.

However, conclusions are usually based only on detected statistically significant associations.

COMPLIANCE WITH ETHICAL STANDARDS

The authors declare that they have no conflict of interest. This article does not contain any studies involving animals or human participants performed by any of the authors.

REFERENCES

1. Ichas, M., *O prirode zhivogo: mekhanizmy i smysl* (On the Nature of the Living: Mechanisms and Meaning), Moscow: Mir, 1994.
2. Hofmann, B., Holm, S., and Iversen, J.-G., Philosophy of science, in *Research Methodology in the Medical and Biological Sciences*, Laake, P., Benestad, H.B., and Olsen, B.R., Eds., Academic Press, Elsevier, 2007, pp. 1–32.
3. *A Dictionary of Epidemiology*, Last, J.M. and Abramson, J.H., Eds., Oxford Univ. Press, 1995, part 554.
4. Epidemiology Glossary, Columbia University Digital Government Research Center, 2009. <http://www.cs.columbia.edu/digigov/LEXING/CDCEPI/gloss.html>. Accessed May 16, 2018.
5. Kundi, M., Causality and the interpretation of epidemiologic evidence, *Environ. Health Perspect.*, 2006, vol. 114, no. 7, pp. 969–974.
6. Ozhegov, S.I. and Shvedova, N.Yu., *Tolkovyi slovar' russkogo yazyka* (Explanatory Dictionary of Russian Language), 4th ed. (extended), Moscow: Azbukovnik, 1999.
7. Yarmonenko, S.P. and Vainson, A.A., *Radiobiologiya cheloveka i zivotnykh* (Radiobiology of Humans and animals), Moscow: Vysshaya Shkola, 2004.
8. Parascandola, M. and Weed, D.L., Causation in epidemiology, *J. Epidemiol. Community Health*, 2001, vol. 55, no. 12, pp. 905–912.
9. Kaufman, J.S. and Poole, C., Looking back on “causal thinking in the health sciences,” *Annu. Rev. Public Health*, 2000, vol. 21, pp. 101–119.
10. Rothman, K.J., Causes, *Am. J. Epidemiol.*, 1976, vol. 104, no. 6, pp. 587–592.
11. Susser, M., What is a cause and how do we know one? A grammar for pragmatic epidemiology, *Am. J. Epidemiol.*, 1991, vol. 133, no. 7, pp. 635–648.
12. Milroy, G., The case of the Egyptian frigate at Liverpool., with remarks on the causation of fevers, &c, *Med. Crit. Psychol. J.*, 1861, vol. 1, no. 4, pp. 552–563.
13. Hill, B.A., The environment and disease: association or causation?, *Proc. R. Soc. Med.*, 1965, vol. 58, pp. 295–300.
14. Phillips, C.V. and Goodman, K.J., Hill's considerations for causal inference, in *Encyclopedia of Epidemiology*, Boslaugh, S., Ed., Two Volume Set, Saint Louis University, SAGE Publications, Inc., 2008, pp. 494–495.
15. *Epidemiologicheskii slovar'* (Epidemiological Dictionary), Last, J.M. and Vlasov, V.V., Eds., Moscow: Otkrytyi institut zdorov'ya v ramkakh proekta “Glo-bus,” 2009.
16. Shkarin, V.V. and Kovalishena, O.V., The etiology of new infections, *Med. Kuzbassa*, 2013, vol. 12, no. 2, pp. 13–21. <http://globalf5.com/Zhurnaly/Medicina/Medicina-v-Kuzbasse/vypusk-2013-2>. Accessed May 16, 2018.
17. Susser, M., *Causal Thinking in the Health Sciences: Concepts and Strategies of Epidemiology*, New York: Oxford Univ. Press, 1973.
18. Gage, S.H., Munafo, M.R., and Davey Smith, G., Causal inference in developmental origins of health and disease (DOHaD) research, *Annu. Rev. Psychol.*, 2016, vol. 67, pp. 567–585.
19. Boice, J.D., Radiation epidemiology and recent paediatric computed tomography studies, *Ann. ICRP*, 2015, vol. 44, no. 1 (suppl.), pp. 236–248.
20. Kotero, A.N., Ushenkova, L.N., and Biryukov, A.P., The issue of the onset of a “New era” in the epidemiology of small doses of radiation (review), *Saratov. Nauchno-Med. Zh.*, 2016, vol. 12, no. 4, pp. 654–662.
21. *UNSCEAR 2012. Report to the General Assembly, with Scientific Annexes*, annex A: *Attributing Health Effects to Ionizing Radiation Exposure and Inferring Risks*, New York, 2015.
22. *UNSCEAR 2013. Report to the General Assembly, with Scientific Annexes*, vol. II, annex B: *Effects of Radiation Exposure of Children*, New York, 2013, pp. 1–268.
23. Pazaitou-Panayiotou, K., Michalakis, K., and Paschke, R., Thyroid cancer in patients with hyperthyroidism, *Horm. Metab. Res.*, 2012, vol. 44, no. 4, pp. 255–262.
24. *UNSCEAR 2000. Report to the General Assembly, with Scientific Annexes*, annex J: *Exposures and Effects of the Chernobyl Accident*, New York, 2000, pp. 451–566.
25. *UNSCEAR 2008. Report to the General Assembly, with Scientific Annexes*, annex D: *Health Effects due to Radiation from the Chernobyl Accident*, New York, 2011, pp. 47–219.
26. Vierra, A., Pollock, J., and Golez, F., *Reading Educational Research*, 3rd. ed., Upper Saddle River, N.J.: Merrill/Prentice Hall, 1992.
27. Cottrell, R. and McKenzie, J.F., *Health Promotion & Education Research Methods: Using the Five Chapter Thesis/Dissertation Model*, 2nd ed., Jones & Bartlett Learning, 2010.
28. Heath, W., *Psychology Research Methods: Connecting Research to Students' Lives*, Cambridge Univ. Press, 2018.
29. Kundi, M., Causality and the interpretation of epidemiologic evidence, *Cien. Saude Colet.*, 2007, vol. 12, no. 2, pp. 419–428.
30. Delarue, N.C., Smoking or health: an early historical perspective, *Ann. R. College Physicians Surgeons Canada*, 1990, vol. 23, no. 6, pp. 431–435.

31. Cunningham R., *Smoke and Mirrors, The Canadian Tobacco War*, Ottawa: Int. Dev. Res. Centre, 1996.
32. Fletcher, R., Fletcher, S., and Vagner, E., *Klinicheskaya epidemiologiya. Osnovy dokazatel'noi meditsiny* (Clinical Epidemiology: Fundamentals of Evidence-Based Medicine), Moscow: Media Sfera, 1998.
33. Vlasov, V.V., *Epidemiologiya: uchebnoe posobie* (Epidemiology: A Tutorial), 2nd ed. (revised), Moscow: GEOTAR-Media, 2006.
34. Evans, A.S., Causation and disease: the Henle–Koch postulates revisited, *Yale J. Biol. Med.*, 1976, vol. 49, pp. 175–195.
35. Doll, R., Tobacco: a medical history, *J. Urban Health*, 1999, vol. 76, no. 3, pp. 289–313.
36. Weed, D.L., Causal criteria and Popperian refutation, in *Causal Inference*, Rothman, K.J., Ed., Massachusetts, USA: Epidemiology Resources Inc., 1988, pp. 15–32.
37. *The Health Consequences of Smoking: A Report of the Surgeon General Rockville*, MD: Office of the Surgeon General, US Public Health Service, 2004.
38. Morabia, A., On the origin of Hill's causal criteria, *Epidemiology*, 1991, vol. 2, no. 5, pp. 367–369.
39. Parascandola, M., Two approaches to etiology: the debate over smoking and lung cancer in the 1950s, *Endeavour*, 2004, vol. 28, no. 2, pp. 81–86.
40. Thygesen, L.C., Andersen, G.S., and Andersen, H., A philosophical analysis of the Hill criteria, *J. Epidemiol. Commun. Health*, 2005, vol. 59, no. 6, pp. 512–516.
41. Clarke, B., *Philosophy of Medicine, The Module for Students*, #Chapter 3: Causation in medicine, London: The University College London (UCL) Department of Science and Technology Studies, 2016. <https://github.com/Brendan-Clarke/Philosophy-of-Medicine/blob/master/Chapter%203.md>. Accessed May 16, 2018.
42. Blackburn, H. and Labarthe, D., Stories from the evolution of guidelines for causal inference in epidemiologic associations: 1953–1965, *Am. J. Epidemiol.*, 2012, vol. 176, no. 12, pp. 1071–1077.
43. *Smoking and Health: Report of the Advisory Committee to the Surgeon General of the Public Health Service Publication no. 1103*, Washington DC: U.S. Department of Health, Education and Welfare, 1964.
44. Rothman, K.J. and Greenland, S., Causation and causal inference in epidemiology, *Am. J. Publ. Health*, 2005, vol. 95, suppl. 1, pp. S144–S150.
45. Evans, D.W., Lucas, N., and Kerry, R., Time, space and form: necessary for causation in health, disease and intervention?, *Med. Health Care Philos.*, 2016, vol. 19, no. 2, pp. 207–213.
46. Hofler, M., The Bradford Hill considerations on causality: a counterfactual perspective, *Emerg. Themes Epidemiol.*, 2005, vol. 2, no. 11. <https://doi.org/10.1186/1742-7622-2-11>
47. Koterov, A.N., Restrictions in the extrapolation of the laws for cells in vitro to radiation medicine, *Med. Radiol. Radiats. Bezop.*, 2009, vol. 54, no. 5, pp. 5–14.
48. *Proc. 2nd Int. Conf. "The Genetics of Aging and Longevity," Moscow, April 22–25, 2012*, Moscow, 2012. Accessed May 16, 2018. http://fund.scienceagainstaging.com/Books/News_aging-genes_all_final_s.pdf.
49. Koterov, A.N., Search problems of human radiation protection in the world of genetics of aging, *Radiats. Biol. Radioecol.*, 2013, vol. 53, no. 5, pp. 487–494.
50. Koterov, A.N., Zharkova, G.P., and Biryukov, A.P., Tandem of radiation epidemiology and radiobiology for radiation protection practices, *Med. Radiol. Radiats. Bezop.*, 2010, vol. 55, no. 4, pp. 55–84.
51. Koterov, A.N. and Biryukov, A.P., Role of radiobiology for radiation epidemiology using for radiation protection, *Int. J. Low Radiat.*, 2010, vol. 7, no. 6, pp. 473–499.
52. Yudin, A.I., *Istoriya i filosofiya nauki: obshchie problemy: uchebnoe posobie* (History and Philosophy of Science. General Problems: A Tutorial), Tambov: Tambov. Gos. Techn. Univ., 2012.
53. Holland, P.W., Statistics and causal inference, *J. Am. Stat. Assoc.*, 1986, vol. 81, no. 396, pp. 945–960.
54. Teichman, J. and Evans, K., *Filosofiya. Rukovodstvo dlya nachinayushchikh* (Philosophy. A Beginner's Guide), Moscow: INFRA-M, Ves' Mir, 1998.
55. Guzelian, P.S., Victoroff, M.S., Halmes, N.C., et al., Evidence-based toxicology: a comprehensive framework for causation, *Hum. Exp. Toxicol.*, 2005, vol. 24, no. 4, pp. 161–201.
56. *Istoriya filosofii: uchebnyk dlya vuzov* (History of Philosophy: A Textbook for Universities), Vasil'ev, V.V., Krotov, A.A., and Bugaya, D.V., Eds., Moscow: Akadem. Proekt, 2005.
57. Coughlin, S.S., *Causal Inference and Scientific Paradigms in Epidemiology*, Bentham E-book, 2010. <https://doi.org/10.2174/97816080518161100101>. <https://ebooks.benthamsience.com/book/9781608051816>. Accessed May 16, 2018.
58. Disease causality (lection), in *Immunology and General Pathology*, vol. II: *General Pathology*, Lection List, Facolta di Medicina e Chirurgia, Universita degli Studi di Pavia (portal <http://nfs.unipv.it/>). http://nfs.unipv.it/nfs/minf/dispense/patgen/lectures/files/disease_causality.html. Accessed May 16, 2018.
59. Hume, D., *Sochineniya* (Works), in 2 vols., 2nd ed. (revised and extended), Moscow: Mysl', 1996.
60. Buchilo, N.F. and Chumakov, A.N., *Filosofiya: Uchebnoe posobie* (Philosophy: A Textbook), 2nd ed. (revised and extended), Moscow: Per Se, 2001.
61. Straud, B., Naturalism and skepticism in the philosophy by Hume, *Epistemol. Filos. Nauki*, 2012, vol. 31, no. 1, pp. 36–51.
62. Mill, D.S., *Sistema logiki sillogisticheskoi i induktivnoi: izlozhenie printsipov dokazatel'stva v svyazi s metodami nauchnogo issledovaniya* (The System of Syllogistic and Inductive Logics: Statement of Principles of Evidence in Connection with the Methods of Scientific Research), 5th ed. (revised and extended), Moscow, 2011.

63. Gay, J., Clinical Epidemiology and Evidence-Based Medicine Glossary: Terminology Specific to Epidemiology, 2005. <http://people.vetmed.wsu.edu/jmgay/courses/GlossEpiTerminology.htm>. Accessed May 16, 2018.
64. Maclure, M., Popperian refutation in epidemiology, *Am. J. Epidemiol.*, 1985, vol. 121, no. 3, pp. 343–350.
65. Weed, D.L. and Trock, B.J., Criticism and the growth of epidemiologic knowledge (Re: ‘Popperian refutation in epidemiology’), *Am. J. Epidemiol.*, 1986, vol. 123, no. 6, pp. 1119–1121.
66. Susser, M., The logic of Sir Karl Popper and the practice of epidemiology, *Am. J. Epidemiol.*, 1986, vol. 124, no. 5, pp. 711–718.
67. Koch, E., Otarola, A., and Kirschbaum, A., A landmark for Popperian epidemiology: refutation of the randomised Aldactone evaluation study, *J. Epidemiol. Commun. Health*, 2005, vol. 59, no. 11, pp. 1000–1006.
68. Hulswit, M., *From Cause to Causation*, Springer Science and Business Media, 2002, Chapter 1: A short history of ‘Causation’. <http://see.library.utoronto.ca/SEED/Vol4-3/Hulswit.htm>. Accessed May 16, 2018.
69. *Filosofiya nauki: Uchebnoe posobie dlya vuzov* (Philosophy of Science: A Textbook for Universities), Troegubov, A.G., Ed., Tula: Tul’sk. Gos. Univ., 2006.
70. *Filosofiya nauki: Uchebnoe posobie dlya vuzov* (Philosophy of Science: A Textbook for Universities), Lebedev, S.A., Ed., Moscow: Akadem. proekt Al’ma Mater, 2007.
71. Lenin, V.I., *Materializm i empiriokrititsizm* (Materialism and Empiriocriticism), vol. 18 of *Polnoe sobranie sochinenii* (The Complete Works), 5th ed., Moscow: Izd. Politicheskoi Literatury, 1968.
72. Parascandola, M., Causes, risks, and probabilities: probabilistic concepts of causation in chronic disease epidemiology, *Prev. Med.*, 2011, vol. 53, nos. 4–5, pp. 232–234.
73. Pearson, K., *The Grammar of Science Paperback*, New York: Cosimo Classics, 2007.
74. Davydovskii, I.V., *Problema prichinnosti v meditsine (etiologiya)* (The Problem of Causality in Medicine (Etiology)), Moscow: Gos. Izd. Meditsinskoi Literatury, 1962.
75. Karhausen, L.R., The logic of causation in epidemiology, *Scand. J. Soc. Med.*, 1996, vol. 24, no. 1, pp. 8–13.
76. Chianelli, M., Bizzarri, G., Todino, V., et al., Laser ablation and 131-iodine: a 24-month pilot study of combined treatment for large toxic nodular goiter, *J. Clin. Endocrinol. Metab.*, 2014, vol. 99, no. 7, pp. E1283–E1286.
77. *ICRP Publication 118, ICRP Statement on Tissue Reactions and Early and Late Effects of Radiation in Normal Tissues and Organs—Threshold Doses for Tissue Reactions in a Radiation Protection Context*, *Annals of the ICRP*, Clement, C.H., Eds., Amsterdam: Elsevier, 2012.
78. Stehbens, W.E., The concept of cause in disease, *J. Chronic. Dis.*, 1985, vol. 38, pp. 947–950.
79. Marshall, B.J. and Warren, J.R., Unidentified curved bacilli in the stomach of patients with gastritis and peptic ulceration, *Lancet*, 1984, vol. 1, no. 8390, pp. 1311–1315.
80. Charlton, B.G., Attribution of causation in epidemiology: chain or mosaic?, *J. Clin. Epidemiol.*, 1996, vol. 49, no. 1, pp. 105–107.
81. Rifkin, E. and Bouwer, E., *The Illusion of Certainty: Health Benefits and Risks*, Boston: MA Springer US, 2007.
82. Rothman, K.J. and Poole, C., A strengthening programme for weak associations, *Int. J. Epidemiol.*, 1988, vol. 17, no. 4, pp. 955–959.
83. Rothman, K.J., What is causation, in *Epidemiology, an Introduction*, Rothman, K.J., Ed., New York: Oxford Univer. Press, 2002, pp. 8–24.
84. Rothman, K.J., Greenland, S., Poole, C., and Lash, T.L., Causation and causal inference, in *Modern Epidemiology*, Rothman, K.J., Greenland, S., and Lash, T.L., Eds., 3rd ed., Philadelphia (PA): Wolters Kluwer, 2008, pp. 5–31.
85. Rothman, K.J., The growing rift between epidemiologists and their data, *Eur. J. Epidemiol.*, 2017, vol. 32, no. 10, pp. 863–865.
86. Hoffmann, K., Heidemann, C., Weikert, C., et al., Estimating the proportion of disease due to classes of sufficient causes, *Am. J. Epidemiol.*, 2006, vol. 163, no. 1, pp. 76–83.
87. Gerstman, B.B., *Epidemiology Kept Simple: An Introduction to Traditional and Modern Epidemiology*, 2nd ed., New York: Wiley-Liss, 2003.
88. Wensink, M., Westendorp, R.G., and Baudisch, A., The causal pie model: an epidemiological method applied to evolutionary biology and ecology, *Ecol. Evol.*, 2014, vol. 4, no. 10, pp. 1924–1930.
89. Pai, M., *Fundamentals of Epidemiology*, Lectures, McGill University, Montreal, Canada, 2014. <https://www.teachepi.org/courses/fundamentals-of-epidemiology/>. Accessed May 16, 2018.
90. Verbeek, J., When work is related to disease, what establishes evidence for a causal relation?, *Saf. Health Work*, 2012, vol. 3, no. 2, pp. 110–116.
91. *Is This Association Causal?*, Lecture, Boston University School of Public Health (according to the used references, no earlier than 2014). http://sphweb.bumc.bu.edu/otlt/MPH-Modules/QuantCore/PH717_Causality/PH717_Causality_print.html. Accessed May 16, 2018.
92. Kim, J., Epidemiology and causation, *Kosin. Med. J.*, 2013, vol. 28, no. 2, pp. 87–97. <https://doi.org/10.7180/kmj.2013.28.2.87>
93. Krieger, N., Epidemiology and the web of causation: has anyone seen the spider?, *Soc. Sci. Med.*, 1994, vol. 39, no. 7, pp. 887–903.
94. Al’bom, A. and Norell, S., *Vvedenie v sovremennuyu epidemiologiyu* (Introduction to Modern Epidemiology), Tallinn, 1996.

95. Briko, N.I., *Epidemiological Approach to the Study of Human Diseases and the Fundamentals of Evidence-Based Medicine*, Lecture, Moscow: Mosk. Med. Akad. im. I.M. Sechenova, 2007. <https://studfiles.net/preview/1350459/page:5/>. Accessed May 16, 2018.
96. Panov, V.G. and Nagrebetskaya, Yu.V., Algebraic interpretation of the two-factor binary theory of sufficient reasons, *Tr. SPIIRAN*, 2013, vol. 26, no. 3, pp. 277–296.
97. Lewis, D., Causation, *J. Philos.*, 1973, vol. 70, no. 17, pp. 556–567.
98. Mackie, J.L., *The Cement of the Universe: A Study of Causation*, Oxford: Clarendon Press, 1974.
99. Olsen, J., What characterises a useful concept of causation in epidemiology?, *Epidemiol. Commun. Health*, 2003, vol. 57, no. 2, pp. 86–88.
100. *Kantserogenez (Carcinogenesis)*, Zaridze, D.G., Ed., Moscow: Meditsina, 2004.
101. *UNSCEAR 2006, Report to the General Assembly, with Scientific Annexes*, annex B; *Epidemiological Evaluation of Cardiovascular Disease and Other Non-Cancer Diseases Following Radiation Exposure*, New York: 2006, pp. 325–383.
102. *UNSCEAR 2010. Report of the United Nations Scientific Committee on the Effects of Atomic Radiation 2010, Fifty-Seventh Session, Includes Scientific Report: Summary of Low-Dose Radiation Effects on Health*, New York, 2011.
103. Koterov, A.N. and Vainson, A.A., Biological and medical effects of radiation with a low linear energy transfer for different dose ranges, *Med. Radiol. Radiats. Bezop.*, 2015, vol. 60, no. 3, pp. 5–31.
104. MacMahon, B., Gene–environment interaction in human disease, *J. Psychiat. Res.*, 1968, vol. 6, suppl. 1, pp. 393–401.
105. *UNSCEAR 2000, Report to the General Assembly, with Scientific Annexes*, annex F: *DNA Repair and Mutagenesis*, New York, 2000, pp. 1–72.
106. *UNSCEAR 2000. Report to the General Assembly, with Scientific Annexes*, annex H: *Combined Effects of Radiation and Other Agents*, New York, 2000, pp. 177–295.
107. Legeza, V.I., Grebenyuk, A.N., and Boyarintsev, V.V., *Kombinirovannye radiatsionnye porazheniya i ikh komponenty (Combined Radiation Injuries and Their Components)*, St. Petersburg: Foliant, 2015.
108. Petin, V.G. and Kim, G.K., *Synergistic Interaction and Cell Responses in Environmental Factors*, New York: Nova Sci. Publ., 2016.
109. Van der Mei, I.A., Simpson, S. Jr., Stankovich, J., and Taylor, B.V., Individual and joint action of environmental factors and risk of MS, *Neurol. Clin.*, 2011, vol. 29, no. 2, pp. 233–255.
110. Liao, S.F., Lee, W.C., Chen, H.C., et al., Baseline human papillomavirus infection, high vaginal parity, and their interaction on cervical cancer risks after a follow-up of more than 10 years, *Cancer Causes Control*, 2012, vol. 23, no. 5, pp. 703–708.
111. Liao, S.F., Yang, H.I., Lee, M.H., et al., Fifteen-year population attributable fractions and causal pies of risk factors for newly developed hepatocellular carcinomas in 11,801 men in Taiwan, *PLoS One*, 2012, vol. 7, no. 4, e34779. <https://doi.org/10.1371/journal.pone.0034779>
112. Lynn, K.S., Lu, C.H., Yang, H.Y., et al., Construction of gene clusters resembling genetic causal mechanisms for common complex disease with an application to young-onset hypertension, *BMC Genomics*, 2013, vol. 14, p. 497. <https://doi.org/10.1186/1471-2164-14-497>
113. Chen, Y.C., Kao, C.F., Lu, M.K., et al., The relationship of family characteristics and bipolar disorder using causal-pie models, *Eur. Psychiatry*, 2014, vol. 29, no. 1, pp. 36–43.
114. Chen, W.C. and Plewig, G., Are demodex mites principal, conspirator, accomplice, witness or bystander in the cause of Rosacea?, *Am. J. Clin. Dermatol.*, 2015, vol. 16, no. 2, pp. 67–72.
115. Johnson, C.Y. and Howards, P.P., Causal pie bingo!, *Epidemiology*, 2013, vol. 24, no. 2, p. 331.
116. Dawber, T.R., Kannel, W.B., Revotskie, N., et al., Some factors associated with the development of coronary heart disease: six years' follow-up experience in the Framingham study, *Am. J. Publ. Health Nat. Health*, 1959, vol. 49, pp. 1349–1356.
117. MacMahon, B., Pugh, T.F., and Ipsen, J., *Epidemiologic Methods*, Boston: Little, Brown and Co., 1960.
118. Pugh, T., *Epidemiology: Principles and Methods*, Boston: Little, Brown and Co., 1970.
119. Cartwright, N., *Nature's Capacities and Their Measurement*, Oxford: Clarendon Press, 1989.
120. Greenland, S. and Robins, J.M., Identifiability, exchangeability, and epidemiological confounding, *Int. J. Epidemiol.*, 1986, vol. 15, no. 3, pp. 413–419.
121. Mangano, J.J., Gould, J.M., Sternglass, E.J., et al., Infant death and childhood cancer reductions after nuclear plant closings in the united states, *Arch. Environ. Health*, 2002, vol. 57, no. 1, pp. 23–31.
122. Mangano, J.J. and Sherman, J.D., Childhood leukaemia near nuclear installations, *Eur. J. Cancer Care (Engl.)*, 2008, vol. 17, no. 4, pp. 416–418.
123. Fecht, S., What can we do about junk science?, *Popular Science*, 21. 05.2014. <https://www.popularmechanics.com/science/health/a10329/what-can-we-do-about-junk-science-16674140>. Accessed May 16, 2018.
124. *COMARE (Committee on Medical Aspects of Radiation in the Environment), 17th Report, Further Consideration of the Incidence of Cancers around the Nuclear Installations at Sellafield and Dounreay*, Produced by Public Health England for the COMARE, 2016.
125. Yablokov, A.V., *Mifo bezopasnosti malykh doz radiatsii: atomnaya mifologiya (The Myth of the Safety of Small Doses of Radiation: Nuclear Mythology)*, Moscow: Tsentr Ekol. Politiki Rossii, OOO Proekt-F, 2002.
126. Stahl, O., Eberhard, J., Jepson, K., et al., The impact of testicular carcinoma and its treatment on sperm

- DNA integrity, *Cancer*, 2004, vol. 100, no. 6, pp. 1137–1144.
127. Chrzanowska, K.H., Gregorek, H., Dembowska-Baginska, B., et al., Nijmegen breakage syndrome (NBS), *Orphanet J. Rare Dis.*, 2012, vol. 7, article 13. <http://www.ordjrd.com/content/7/1/13>. Accessed May 16, 2018.
 128. Li, F.P., Dreyfus, M.G., and Antman, K.H., Asbestos-contaminated nappies and familial mesothelioma, *Lancet*, 1989, vol. 1, no. 8643, pp. 909–910.
 129. Li, F.P., Lokich, J., Lapey, J., et al., Familial mesothelioma after intense asbestos exposure at home, *JAMA*, 1978, vol. 240, no. 5, p. 467.
 130. Weed, D.L., Beyond black box epidemiology, *Am. J. Publ. Health*, 1998, vol. 88, no. 1, pp. 12–14.
 131. Savitz, D.A., In defense of black box epidemiology, *Epidemiology*, 1994, vol. 5, no. 5, pp. 550–552.
 132. Mineyko, A. and Kirton, A., The black box of perinatal ischemic stroke pathogenesis, *J. Child.*, 2011, vol. 26, no. 9, pp. 1154–1162.
 133. Yerushalmy, J. and Palmer, C.E., On the methodology of investigations of etiologic factors in chronic diseases, *J. Chronic. Dis.*, 1959, vol. 10, no. 1, pp. 27–40.
 134. Susser, M. and Susser, E., Choosing a future for epidemiology: II. From black box to Chinese boxes and eco-epidemiology, *Am. J. Publ. Health*, 1996, vol. 86, no. 5, pp. 674–647.
 135. Poole, C. and Rothman, K.J., Epidemiological science and public health policy, *J. Clin. Epidemiol.*, 1990, vol. 43, no. 11, pp. 1270–1271.
 136. Kleinbaum, D.G., Kupper, L.L., and Morgenstern, H.L., *Epidemiologic Research: Principles and Quantitative Methods*, Belmont, CA: Lifetime Learning Publications, 1982.
 137. Kuhn, T.S., Objectivity, value judgment, and theory choice, in *The Essential Tension*, Kuhn, T.S., Ed., Chicago: Univ. Chicago Press, 1977, pp. 320–343.
 138. Holman, C.D., Arnold-Reed, D.E., de Klerk, N., et al., A psychometric experiment in causal inference to estimate evidential weights used by epidemiologists, *Epidemiology*, 2001, vol. 12, no. 2, pp. 246–255.
 139. Jenkins-Smith, H.C., Silva, C.L., and Murray, C., Belief about radiation: scientists, the public and public policy, *Health Phys.*, 2009, vol. 97, no. 5, pp. 519–527.
 140. Tranoy, K.E., Science and ethics. Some of the main principles and problems, in *The Moral Import of Science: Essays on Normative Theory, Scientific Activity and Wittgenstein*, Jones, A.J.I., Ed., Bergen: Sigma Ferlag, 1988, pp. 111–136.
 141. EEA (European Environment Agency). *Late Lessons from Early Warnings: the Precautionary Principle 1896–2000*, Environmental Issue Report no. 22, Copenhagen: European Environment Agency, 2001.
 142. COMEST. *The Precautionary Principle*, World Commission on the Ethics of Scientific Knowledge and Technology, Ed., Paris: United Nations Educational, Scientific and Cultural Organization, 2005.
 143. Health Council of the Netherlands, *Prudent Precaution, Publication no. 2008/18E*, The Hague: Health Council of the Netherlands, 2008.
 144. Gray, J.S., Statistics and the precautionary principle, *Mar. Pollut. Bull.*, 1990, vol. 21, no. 4, pp. 174–176.
 145. Johnston, P. and Simmonds, M., Precautionary principle, *Mar. Pollut. Bull.*, 1990, vol. 21, no. 8, p. 402.
 146. Johnston, P. and Simmonds, M., Green light for precautionary science, *New Sci.*, 1991, vol. 132, no. 1780, p. 4.
 147. Tallacchini, M., Before and beyond the precautionary principle: epistemology of uncertainty in science and law, *Toxicol. Appl. Pharmacol.*, 2005, vol. 207, no. 2, pp. 645–651.
 148. Holm, S. and Takala, T., High hopes and automatic escalators: a critique of some new arguments in bioethics, *J. Med. Ethics*, 2007, vol. 33, no. 1, pp. 1–4.
 149. Stabell, E.D., Constraints on the precautionary principle and the problem of uncertainty, *Am. J. Bioeth.*, 2017, vol. 17, no. 3, pp. 56–57.
 150. Stirling, A. and Coburn, J., From CBA to precautionary appraisal: practical responses to intractable problems, *Hastings Cent. Rep.*, 2018, vol. 48, suppl. 1, pp. S78–S87.
 151. Crichton, M., *Aliens Cause Global Warming. Caltech Michelin Lecture, 2003* (various Web sources, e.g., Sceptical Science (non-profit science education organization). https://www.skepticalscience.com/Crichton_Aliens_Cause_Global_Warming.html. Accessed May 16, 2018.
 152. Morabia, A., History of the modern epidemiological concept of confounding, *J. Epidemiol. Commun. Health*, 2011, vol. 65, pp. 297–300.
 153. Kornysheva, E.A., Platonov, D.Yu., Rodionov, A.A., and Shabashov, A.E., *Epidemiologiya i statistika kak instrumenty dokazatel'noi meditsiny* (Epidemiology and Statistics as Tools for Evidence-Based Medicine), 2nd ed. (revised and extended), Tver, 2009.
 154. Weinberg, C.R., Toward a clearer definition of confounding, *Am. J. Epidemiol.*, 1993, vol. 137, no. 1, pp. 1–8.
 155. Haneuse, S., Distinguishing selection bias and confounding bias in comparative effectiveness research, *Med. Care*, 2016, vol. 54, no. 4, pp. e23–e29.
 156. Fisher, R., Cigarettes, cancer, and statistics, *Centennial Rev.*, 1958, vol. 2, pp. 151–166.
 157. Bhopal, R.S., *Concepts of Epidemiology: An Integrated Introduction to the Ideas, Theories, Principles and Methods of Epidemiology*, Oxford: Univ. Press, 2002.
 158. Smith, G.D., Phillips, A.N., and Neaton, J.D., Smoking as 'independent' risk factor for suicide: illustration of an artifact from observational epidemiology?, *Lancet*, 1992, vol. 340, no. 8821, pp. 709–712.
 159. Axelson, O., Confounding from smoking in occupational epidemiology, *Br. J. Ind. Med.*, 1989, vol. 46, no. 8, pp. 505–507.

160. Galarraga, V. and Boffetta, P., Coffee drinking and risk of lung cancer—meta-analysis, *Cancer Epidemiol. Biomarkers Prev.*, 2016, vol. 25, no. 6, pp. 951–957.
161. Gordis, L. and Gold, E.B., Epidemiology and etiology of pancreatic cancer, in *The Pancreas: Biology, Pathobiology, and Disease*, Way Liang W. Go, et al., Eds., 2nd ed., New York: Raven Press, Ltd, 1993, pp. 837–855.
162. Ciccozzi, M., Menga, R., Ricci, G., et al., Critical review of sham surgery clinical trials: confounding factors analysis, *Ann. Med. Surg. (Lond.)*, 2016, vol. 29, no. 12, pp. 21–26.
163. Stark, C.R. and Mantel, N., Effects of maternal age and birth order on the risk of mongolism and leukemia, *J. Natl. Cancer Inst.*, 1966, vol. 37, no. 5, pp. 687–698.
164. Petitti, D.B., Perlman, J.A., and Sidney, S., Noncontraceptive estrogens and mortality: long-term follow-up of women in the Walnut Creek Study, *Obstet. Gynecol.*, 1987, vol. 70, no. 3, pp. 289–293.
165. Kallen, B., The problem of confounding in studies of the effect of maternal drug use on pregnancy outcome, *Obstet. Gynecol. Int.*, 2012, vol. 2012, art. 148616. <https://doi.org/10.1155/2012/148616>
166. Bruning, O., Rodenburg, W., Wackers, P.F., et al., Confounding factors in the transcriptome analysis of an in-vivo exposure experiment, *PLoS One*, 2016, vol. 11, no. 1. e0145252. <https://doi.org/10.1371/journal.pone.0145252>
167. Weed, D.L., Historical roots of the healthy worker effect, *J. Occup. Med.*, 1986, vol. 28, no. 5, pp. 343–347.
168. McGeoghegan, D. and Binks, K., The mortality and cancer morbidity experience of employees at the Chapelcross plant of British Nuclear Fuels plc, 1955–1995, *J. Radiol. Prot.*, 2001, vol. 21, no. 3, pp. 221–250.
169. Telle-Lamberton, M., Samson, E., Caer, S., et al., External radiation exposure and mortality in a cohort of French nuclear workers, *Occup. Environ. Med.*, 2007, vol. 64, no. 10, pp. 694–700.
170. Muirhead, C.R., Goodill, A.A., Haylock, R.G.E., et al., Occupational radiation exposure and mortality: second analysis of the National Registry for Radiation Workers, *J. Radiol. Prot.*, 1999, vol. 19, no. 1, pp. 3–26.
171. Preston, D.L., Cigarette smoking and radiation dose in the Life Span Study, *RERF Update*, 1999, vol. 10, no. 2, p. 9.
172. UNSCEAR 2006. *Report to the General Assembly, with Scientific Annexes*, annex A: *Epidemiological Studies of Radiation and Cancer*, New York: 2008, pp. 17–322.
173. COMARE (Committee on Medical Aspects of Radiation in the Environment), 7th Rep.: *Parents Occupationally Exposed to Radiation Prior to the Conception of Their Children. A Review of the Evidence Concerning the Incidence of Cancer in Their Children*, Crown, Ed., Produced by the National Radiological Protection Board, 2002.
174. Sanders, C.L. and Scott, B.R., Smoking and hormesis as confounding factors in radiation pulmonary carcinogenesis, *Dose-Response*, 2006, vol. 6, no. 1, pp. 53–79.
175. Kudo, S., Ishida, J., Yoshimoto, K., et al., The adjustment effects of confounding factors on radiation risk estimates: findings from a Japanese epidemiological study on low-dose radiation effects (J-EPISODE), *J. Mol. Genet. Med.*, 2017, vol. 11, no. 3. <https://doi.org/10.4172/1747-0862.1000275>
176. Koterov, A.N. and Biryukov, A.P., Children of participants of liquidation of consequences of the Chernobyl nuclear power plant. 2. The frequency of deviations and pathologies and their relationship to the non-radiation factors, *Med. Radiol. Radiats. Bezop.*, 2012, vol. 57, no. 2, pp. 51–77.
177. Koterov, A.N. and Biryukov, A.P., The possibility of determining of anomalies and pathologies in the offspring of liquidators of Chernobyl accident by non-radiation factors, *Int. J. Low Radiat. (Paris)*, 2011, vol. 4, no. 8, pp. 256–312.
178. Sonne, J.C., The varying behaviors of fathers in the prenatal experience of the unborn: protecting, loving and “welcoming with arms wide open” vs. ignoring, unloving, competitive, abusive, abortion minding or aborting, *J. Prenat. Perinat. Psychol. Health*, 2005, no. 4, pp. 319–340.
179. McNeil, T.F., Schubert, E.W., Cantor-Graae, E., et al., Unwanted pregnancy as a risk factor for offspring schizophrenia-spectrum and affective disorders in adulthood: a prospective high-risk study, *Psychol. Med.*, 2009, vol. 39, no. 6, pp. 957–965.
180. IPPNW (International Physicians for the Prevention of Nuclear War), *Critical analysis of the UNSCEAR report*, German Affiliate of the International Physicians for the Prevention of Nuclear War, 2014.
181. UNSCEAR 2015. *Report to the General Assembly, with Scientific Annexes, Developments since the 2013 UNSCEAR Report on the Levels and Effects of Radiation Exposure Due to the Nuclear Accident Following the Great East-Japan Earthquake and Tsunami*, New York, 2015.
182. Guzeev, G.G. and Kalabushkin, B.A., Genetic consequences of the Chernobyl accident. Monitoring of congenital malformations of newborns in the Kaluga oblast, *Radiats. Biol. Radioecol.*, 1995, vol. 35, no. 5, pp. 640–646.
183. Ron, E., Lubin, J., and Schneider, A.B., Thyroid cancer incidence, *Nature*, 1992, vol. 360, no. 6400, p. 113.
184. Verkooijen, H.M., Fioretta, G., Pache, J.-C., et al., Diagnostic changes as a reason for the increase in papillary thyroid cancer incidence in Geneva, Switzerland, *Cancer Causes Control*, 2003, vol. 14, no. 1, pp. 13–17.
185. Morris, L.G., Sikora, A.G., Tosteson, T.D., and Davies, L., The increasing incidence of thyroid cancer: the influence of access to care, *Thyroid*, 2013, vol. 23, no. 7, pp. 885–891.
186. Tsuda, T., Tokinobu, A., Yamamoto, E., and Suzuki, E., Thyroid cancer detection by ultrasound among resi-

- dents ages 18 years and younger in Fukushima, Japan: 2011 to 2014, *Epidemiology*, 2016, vol. 27, no. 3, pp. 316–322.
187. Kaiser, J.C., Jacob, P., Blettner, M., and Vavilov, S., Screening effects in risk studies of thyroid cancer after the Chernobyl accident, *Radiat. Environ. Biophys.*, 2009, vol. 48, no. 2, pp. 169–179.
188. Rozhko, A.V., Masyakin, V.B., Nadyrov, E.A., and Okeanov, A.E., The role of the screening effect in the assessment of the results of a cohort study of thyroid pathology, *Med. Radiol. Radiats. Bezop.*, 2010, vol. 55, no. 1, pp. 19–23.
189. Belfiore, A., Russo, D., Vigneri, R., and Filetti, S., Graves' disease, thyroid nodules and thyroid cancer, *Clin. Endocrinol.*, 2001, vol. 55, no. 6, pp. 711–718.
190. Allodji, R.S., Schwartz, B., Veres, C., et al., Risk of subsequent leukemia after a solid tumor in childhood: impact of bone marrow radiation therapy and chemotherapy, *Int. J. Radiat. Oncol. Biol. Phys.*, 2015, vol. 93, no. 3, pp. 658–667.
191. UNSCEAR 1977. *Report to the General Assembly, with Scientific Annexes*, annex G: *Radiation Carcinogenesis in Man*, New York, 1977, pp. 361–423.
192. Dugan, L.C. and Bedford, J.S., Are chromosomal instabilities induced by exposure of cultured normal human cells to low- or high-LET radiation?, *Radiat. Res.*, 2003, vol. 159, no. 3, pp. 301–311.
193. Koterov, A.N., Genomic instability at exposure of low dose radiation with low LET. Mythical mechanism of unproved carcinogenic effects, *Int. J. Low Radiat.*, 2005, vol. 1, no. 4, pp. 376–451.
194. Koterov, A.N., The lack of evidence of genomic instability after exposure of cells without obvious defects and the organism ex utero to small doses of low-LET radiation, *Radiats. Biol. Radioecol.*, 2006, vol. 46, no. 5, pp. 585–596.
195. Rubanovich, A.V. and Khromov-Borisov, N.N., Theoretical analysis of the indices of the predictive efficiency of binary genetic tests, *Ekol. Genet.*, 2013, vol. 11, no. 1, pp. 77–90.

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