

1 Articles

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3 *Call of Duty at the Frontier of Research:*

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6 *Normative Epistemology for High-Risk/High-Gain Studies of*
7 *Deep Brain Stimulation*

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11 QA MERLIN BITTLINGER

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13 **Abstract:** Research participants are entitled to many rights that may easily come into
14 conflict. The most important ones are that researchers respect their autonomy as persons
15 and act on the principles of beneficence, nonmaleficence, and justice. Since 2014, research
16 subjects from numerous states in the United States of America also have a legal “right to
17 try” that allows them, under certain circumstances, to receive experimental (i.e., prelimi-
18 narily tested) interventions, including medical devices, before official approval from the
19 United States Food and Drug Administration. In the context of experimental interven-
20 tions, such as deep brain stimulation (DBS) for Alzheimer’s disease, this article argues
21 that research participants ought never to have a legal “right to try” without a correspond-
22 ing “right to be sure.” The latter refers to external epistemic justification construed in
23 terms of *reliance on reliable evidence*. This article demonstrates that the mere complexity of
24 intervention ensembles, as in the case of DBS for Alzheimer’s disease which serves as a
25 paradigm example, illustrate how unanswered and/or unasked open questions give rise to
26 a “combinatorial explosion” of uncertainties that require epistemic responses that no
27 single research team alone is likely able to provide. From this assessment, several epis-
28 temic asymmetrical relations between researchers and participants are developed. By elu-
29 cidating these epistemic asymmetries, this article unravels the reasons why open science,
30 transparent exhaustive data reporting, preregistration, and continued constant critical
appraisal via pre- and postpublication peer review are not scientific virtues of moral
excellence but rather ordinary obligations of the scientific work routine required to
increase *reliability* and *strength of evidence*.

31 **Keywords:** neuroethics; deep brain stimulation; clinical translation; Alzheimer’s disease;
32 uncertainty; epistemic obligation; social epistemology; culpable ignorance

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34 The translation of insight from the basic sciences into clinical practice is complex,
35 tedious, and expensive. Alzheimer’s disease is a paradigm case. Despite enor-
36 mous research efforts, 99.6% of novel, potential therapeutic agents failed at some
37 point during the clinical translation process between 1990 and 2014.¹ Key players
38 such as Pfizer are already leaving the arena of pharmacological Alzheimer’s dis-
39 ease research.² As such setbacks eventually concern the translation of findings
40 from the brain sciences into clinical medicine, this raises particularly important
41 questions for clinical neuroethics. The translational difficulties are tremendous
42 and the complexity vexing. In consequence, professionals from all areas involved—
43 be they scientists, funding agencies, science journalists, or research ethicists—may
44 feel lost in the maze at some point.³

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1 The argument put forward here makes the point that the epistemic duties of
2 scientists play a crucial role in the translational process from bench to bedside.
3 I attempt to show that doing exploratory research obliges researchers to take
4 advantage of the open science framework. This requires fully transparent, system-
5 atic, and comprehensive publication of all information. Furthermore, I argue that
6 beyond independent prepublication peer review, the scientific community has a
7 duty to engage in meticulous postpublication review of any exploratory high-
8 risk/high-gain research. I will do so by drawing from the example presented by
9 the particular difficulties that arise in the context of “first-in-human” studies
10 investigating deep brain stimulation (DBS) as novel, experimental intervention for
11 new medical indications such as Alzheimer’s disease. In this context, the scientific
12 community as a whole has a *collective epistemic duty* to maximize the evidence, the
13 strength of the evidence, and its reliability, and to use all means available to do so.

14 It is likely that this may strike some readers as trivial and odd at the same time.
15 From a scientist’s perspective, it may seem redundant to call for evidence as the
16 basis of belief formation. Is not the endeavor of science defined as striving for
17 knowledge? In addition, it may also appear to scientists strangely old fashioned to
18 prescribe epistemic obligations at all, and worse, in the language of duties rather
19 than of the virtues of scientific excellence. However, this may immediately change
20 if we introduce some important conceptual distinctions.

21 In epistemology, the “ethics of (secular) belief” captures the idea that there are
22 *epistemic* duties with regard to what one believes. According to the evidentialist
23 stance, one ought to maintain and only to maintain beliefs for which one has suf-
24 ficient reasons.⁴ In this context, “sufficient reasons” means very roughly that, were
25 the evidence that constitutes the reasons entirely true, then the belief would be
26 beyond reasonable doubt when judged by the established standards and criteria
27 for good, credible evidence. In contrast, the pragmatist stance maintains that one
28 should hold those beliefs that maximize pragmatic values such as practical utility,
29 and that we are sometimes positively obliged to form beliefs on insufficient evi-
30 dence. Now, after having made this distinction, scientists may have very different
31 inclinations to either side. Striving for practical utility seems as much a valid sci-
32 entific value as striving for good evidence, and both pragmatist and evidentialist
33 positions are abundant among scientists, research ethicists, and philosophers of
34 science. Therefore, further distinctions are required to justify the evidentialist
35 claims.

36 The thesis that I am going to defend is that scientists ought to prioritize the
37 value of the *strength of evidence* for their beliefs over the value of the *utility of their*
38 *beliefs* when performing high-risk/high gain clinical research involving invasive
39 neurosurgical procedures. I defend this thesis on (normative) epistemic grounds
40 that are largely independent from ethical concerns about potential hazards but,
41 certainly, hazards would constitute additional reasons for concern.⁵

42 In stark contrast to scientists, research participants have no corresponding epis-
43 temic duty to prefer evidence in favor of utility with regard to their belief system,
44 and this may be true for other stakeholders involved as well. Scientists play a
45 particular role in translational research, and this role gives rise to the particular
46 epistemic duty at stake. Against the pragmatist, I argue that scientists are never
47 positively obliged to form beliefs about research on insufficient evidence. In the
48 spirit of the evidentialist, I argue that researchers are often positively obliged to
49 form beliefs on sufficient evidence. However, I drop the universal quantification

1 “always,” as this makes the claim unnecessarily strict. Rather, there are specific
2 conditions under which scientists are strictly obliged to seek further support to
3 strengthen the available evidence.

4 Careful reflection about the social role of the scientist within clinical translation
5 will sharpen scientists’ understanding of their epistemic duties; that is, in techni-
6 cal philosophical terms, “practice-generated entitlements to expect.”⁶ The basic
7 idea is that when scientists realize how much research participants depend epis-
8 temically on the integrity of critical appraisal through the science community, they
9 *ought* to be motivated to accept the normative thesis. That is, to acknowledge the
10 epistemic duty to apply all means and efforts to maximize the strength of evidence
11 and to preempt the hazards and liabilities of *culpable ignorance*.

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Case Study: Ms. Metis is Diagnosed with Alzheimer’s Disease

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16 For illustrative purposes, I will begin with casuistry to set out the general princi-
17 ples that rule the particular case of high-risk/high-gain research. I present the
18 fictitious Ms. Metis, a 63-year-old woman recently diagnosed with probable
19 Alzheimer’s disease with mild cognitive deficits. Terrified by the diagnosis,
20 Ms. Metis remembers her childhood and her experiences as a young woman when
21 she learned from her grandmother what suffering from the relentless progression
22 of neurodegeneration can mean. Ms. Metis is especially frightened because she
23 does not want to burden her family. Her only son is a young father starting his
24 own family and her husband is not in the best of health. Her physician explains
25 that thus far no cure exists. There are only mildly effective pharmacological treat-
26 ments addressing symptoms, and they are not disease modifying. However, there
27 is a novel, innovative brain surgery that seems promising. If Ms. Metis accepts the
28 risks associated, despite that this procedure is in a very early, experimental
29 research phase, she may be eligible to participate in a “first-in-human” trial of DBS
30 for Alzheimer’s disease. Ms. Metis worked as an engineer before the diagnosis
31 and strongly believes in the technological advances of science and medicine.
32 Throughout her life she has taken pride in governing and disciplining herself
33 through the use of reason. Her values are such that she despises idleness and inac-
34 tion. She makes clear to her physician and her family that either she becomes part
35 of that trial or she may choose suicide.⁷

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The Ethical Dimension and Regulatory Oversight

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39 Ms. Metis’s situation is complex and uncertain because of multiple factors. There
40 is not yet an available effective therapy. There is an expected disease burden of
41 great magnitude and probability. There is an ambiguous signal evoked by the
42 research context: caution because of the preliminary status. but perhaps also
43 implicit signs of hope that research is conducted for a purpose, and a tacit assump-
44 tion that even a very small chance is better than no chance at all. The epistemic
45 challenge for Ms. Metis is to avoid severely underestimating the risks or overesti-
46 mating the benefits or both (therapeutic misestimation).⁸

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Avoiding therapeutic misestimation involves acknowledging that despite the
lack of a medical remedy and the high disease burden, additional significant haz-
ards are a possible outcome of “first-in-human” trials of interventions that have
been only preliminarily tested. In truly innovative research, there is only scarce

1 and indirect prior knowledge to inform volunteers in early clinical trials about
2 evidence-based information. Therefore, potential risks may be unknown and indi-
3 vidual medical benefits may be uncertain. Because informed consent is a key cor-
4 nerstone of human research, lack of evidence-based information raises ethical
5 perils by conceptual necessity in “first-in-human” research. As a consequence, the
6 question of regulatory oversight of such research has a conspicuous ethical dimen-
7 sion. Some ethical commentators hold that DBS research using new targets and
8 novel indications should be regarded merely as clinical innovation,⁹ whereas oth-
9 ers criticize this proposal as “neurosurgical exceptionalism.”¹⁰ In consequence, the
10 question of which ethical requirements need to be fulfilled for “first-in-human”
11 DBS research is still open in clinical neuroethics. To cover a broader spectrum of
12 these ethical dimensions, close attention to the context of DBS as a complex inter-
13 vention ensemble is necessary. In particular, a close look at the specific hazards of
14 “first-in-human” studies within the clinical translation process is needed.

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Complex Intervention Ensembles

18 Biology is becoming an increasingly complex information science¹¹ and the trans-
19 lational process itself is an information maximization process.¹² Information, in
20 turn, is best understood as any “difference that makes a difference”¹³ and can be
21 defined as any pattern of data that resolve uncertainty. Therefore, the goal of clini-
22 cal translation is to minimize uncertainty.

23 At the end of the translation process there is not only a new product to sell,
24 there is also uncertainty removed as to whether the new product or intervention
25 is sufficiently effective and safe.¹⁴ Because high-quality clinical research requires
26 expensive resources, as well as voluntary and informed human research subject
27 participation, any scientific study that does not reduce uncertainty about either
28 efficacy, safety, or both, is futile and questionable from an ethical perspective.¹⁵
29 As such, it seems reasonable to assume that clinical translation is also a process
30 of ambiguity aversion. If a novel research question is assessed in a study to
31 resolve genuine uncertainty, then the study should ideally be designed to give a
32 definite answer to that question, not a vague one. It seems *prima facie* better to
33 have a chain of very definite answers to smaller but feasible questions than a
34 sequence of rather vague answers to the next great “breakthrough.” This view
35 complements the very clear distinction between exploratory research and confir-
36 matory research,¹⁶ as what counts as a definite answer may vary in each of these
37 different contexts. In addition, it also requires acknowledging the complexity of
38 any effect brought about by DBS. Even if it is perhaps not necessary to fully
39 understand the mechanism of action of an intervention in “first-in-human”
40 research, there is always a complex interaction of different factors that needs to
41 be taken into account.

42 For DBS, the “intervention ensemble”¹⁷ consists at least of the brain target, the
43 stimulation parameter, and the presumed aberrant brain circuits or brain function
44 as affected by the disease pathology. Therefore, the relevant research question is
45 necessarily quite intricate. In the case of Ms. Metis, the relevant question with
46 regard to what is in her best interest as asked from a detached observer’s position
47 is arguably the following: What is the expected probability that performing DBS
48 in some explicitly defined patient cohort *C* with the specific stimulation parameter
49 settings *S* applied to the particular brain target *T* will effect some physiological

1 change, which is with known probability P_1 positively correlated with a clinically
2 relevant outcome O and which is with known probability P_2 not positively corre-
AQ1 lated with any significant hazards, burdens, or harms that outweigh O ?¹⁸

4 Ideally, the best parameter setting S , the best brain target T and both the proba-
5 bilities P_1 for benefits and P_2 for hazards are well studied in some relevantly simi-
6 lar cohort C of research participants. Typically, in “first-in-human” studies, the
7 latter cohort is some animal model of the respective disease that is investigated.
8 Given this ideal scenario, the uncertainty reduces to the question of whether the
9 effects that hold for cohort C also hold for the cohort of Ms. Metis and her fellow
10 participants. This picture of transferring an effect from one cohort to another is at
11 the heart of the metaphorical content conveyed by “clinical translation.”

12 However, there are several practical complications to this picture. For example,
13 review of ethically relevant questions shows that, in the case of DBS for Alzheimer’s
14 disease, there has been an absence of empirical evidence from preclinical studies
15 prior to “first-in-human studies.”^{19,20} Certainly, the translational gap from rodents
16 to humans is huge,²¹ but there are important yet more modest questions that could
17 be validly examined by high-quality translational animal studies.²² An example is
18 the question of how different sets of stimulation parameters interact with the dis-
19 ease pathology of various distinct animal models, and which stimulation param-
20 eters seem optimal to reach beneficial effects without potential harmful side effects
21 (therapeutic window). Also, the choice of brain target could have been explored
22 antecedently in animal models of Alzheimer’s disease. This information would
23 have been valuable to inform “first-in-human” studies, which were explored
24 using three competing brain targets.^{23,24,25} However, directly relevant studies in
25 mice came only after the “first-in-human” studies.²⁶

26 The lack of prior probabilities derived from animal models has some very seri-
27 ous methodological repercussions that are difficult to spot for scientists them-
28 selves, and are all the more perplexing for research participants in the situation of
29 Ms. Metis.

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31

Exploratory Research

32

33 Lack of accurate and reliable information about prior probabilities of relevant fac-
34 tors is a threat in exploratory research.²⁷ One reason is that it turns a conditional
35 probability into a joint probability. Because of Bayes’ theorem, the conditional
36 probability would be the probability that the desired outcome O may also be
37 observed in another cohort C of research participants; that is, participants of a dif-
38 ferent kind or species, *given the known probabilities* for all other relevant factors
39 such as the certainty about having selected a potentially suitable brain target,
40 stimulation parameter, or disease stage.

41 If these factors cannot be stipulated as *given*, the threat of a combinatorial explo-
42 sion of uncertainties arises. The joint probability of independent events is multi-
43 plicative and Bayes’ theorem does not apply.

44 For illustration, we can simplify the scenario and limit the scope of relevant fac-
45 tors to the number 10 with binary outcomes (yes/no). Then, if DBS experts would
46 be able to estimate the prior probabilities that for each of the 10 relevant factors, a
47 suitable choice can be made with 80% certainty, then the joint probability is calcu-
48 lated as 0.8 to the power of 10; that is multiplying 0.8 10 times. This results in an
49 estimated chance to observe the desired outcome of only 11%, and, respectively,

1 an 89% chance to fail.²⁸ However, if we raise the estimated probability via sound
2 preclinical research to 0.95, then the expected 89% chance to fail turns into a 60%
3 chance of success (0.95 to the power of 10). Although these numbers are fictitious,
4 this clearly calls for rigorous confirmation of exploratory preclinical research using
5 95% or higher confidence intervals to exclude “false positives.”²⁹ It is an important
6 step in risk mitigation to reduce uncertainty about the estimated probabilities of
7 relevant factors and, by doing so, to reduce the risk of “uncertainty blindness.”
8

9 **Risk and “Uncertainty Blindness”**

10
11 According to the United States Food and Drug Administration (FDA) “risk is
12 defined as the combination of the probability of occurrence of harm and the
13 severity of that harm.”³⁰ This definition is useful to describe the situation of
14 Ms. Metis, but it is also defective in an important way. It only applies to known
15 risks. That is, to risks that can be assessed on the basis of already available
16 empirical frequencies. Such empirical frequencies can require sufficiently large
17 numbers of relevant observations to reach statistical reliable estimates for the
18 probability of *occurrences*. Therefore, it seems that risks according to the FDA,
19 can, by definition, not be estimated for “first-in-human” research, in which no
20 prior observations exist.

21 In the case of Ms. Metis, it is obviously not in her best interest to have an exact
22 estimate of potential harms *ex post facto*. She probably wants to know the relevant
23 risk *prior* to undergoing DBS for a novel medical indication such as Alzheimer’s
24 disease in a “first-in-human” research context. In addition, Ms. Metis may also
25 want to take the degree of uncertainty of the estimation into account. According
26 to this, risk can then be defined as the combination of three factors: (1) the esti-
27 mated magnitude of potential harmful consequences, (2) the estimated probabili-
28 ty of the occurrence of any of these potential harmful consequences, and (3) the
29 strength of evidence that justifies arriving at these estimates in a scientific, valid,
30 and reliable way.

31 Decision theory is an established scientific field that studies risk taking under
32 uncertainty and informs complex and difficult questions ranging from financial
33 investments to policymaking and disaster management, but is surprisingly sel-
34 dom used in discussion of “first-in-human” studies of medical interventions.^{31,32}
35 Instead the discussion is focused primarily on the low probabilities of benefits,
36 which treats the probabilities as fixed (e.g., a random variable, where the mathe-
37 matically expected value is the measure of potential benefits or harm [payoff] and
38 the standard deviation is the measure of uncertainty of that benefit or harm to
39 occur). In this simplistic picture, there are many different strategies for making as
40 rational a decision as possible, depending on risk preference such as risk aversion
41 or risk taking. One option is the optimistic strategy of maximizing the likelihood
42 that the *best possible* outcome is as good as possible (risk taker). Another option is
43 to maximize the likelihood that the *worst possible* outcome is as good as possible
44 (risk avoider). It may seem appealing to assume that both researcher and partici-
45 pants who decide to play an active part in “first-in-human” research tend to be
46 risk takers, who want to stay optimistic despite any unfavorable circumstances
47 (see Table 1). Accordingly, risk-averse researchers and participants would seem to
48 avoid “first-in-human” trials right away. Whereas the latter is probably true, the
49 former assumption about risk takers is elusive.

1 **Table 1.** Risk Matrix: A Non-Probability-Based Payoff Table Depicting Therapeutic Mis-
 2 estimation Caused by “Uncertainty Blindness”

		Research participation	
		Yes	No
7 Scenario	Best possible	Some benefit compared with TAU	TAU
8	Most likely	Research burden + effect of TAU	TAU
9	Worst possible	Some harm compared with TAU	TAU

10
 11 If the estimated payoffs were accurate and reliable, risk takers would maximize the best possible sce-
 12 nario that is expected from research participation; that is, some benefit compared with treatment as
 13 usual (TAU), whereas risk-averse candidates would minimize the worst outcome (TAU and nonpartic-
 14 ipation). As alluring as this picture is, it neglects the uncertainty that is associated with estimating
 15 the probability of best and worst possible outcomes or the most likely outcome. It entirely omits any
 16 estimate of the strength of evidence.

17
 18 The whole account discussed so far has an important limitation. It entirely
 19 ignores the *unknown* probabilities of any relevant events that are typical for “first-
 20 in-human” research, and neglects the *uncertainty* of any estimation of relevant
 21 probabilities.³³ To remedy this, the strength of evidence for all relevant factors that
 22 influence the probability of worst or best possible scenarios needs to be taken into
 23 account. A further implication is that factors that drive the uncertainty of the prob-
 24 ability of some potential risk factor need to be explicitly modeled in order to deter-
 25 mine an adequate estimate for the risk. Factors that drive uncertainty are the lack
 26 of reliable empirical evidence and the plausible existence of unknown relevant
 27 factors that influence the probability of the desired study outcome. An example of
 28 the lack of reliable evidence is sparse preclinical data prior to “first-in-human” tri-
 29 als.³⁴ The existence of yet-unknown relevant factors is the more plausible the less
 30 is known about the mechanistic explanations of the indication; for example,
 31 Alzheimer’s disease, and the less is known about the mechanism of action of the
 32 intervention. To be clear, a mechanistic understanding or a lack thereof does not
 33 per se affect the likelihood of a favorable or unfavorable outcome. But the lack of
 34 a mechanistic understanding may increase uncertainty as to whether there are any
 35 relevant factors that have been overlooked while planning a “first-in-human”
 36 study.

37
 38 **Epistemic Asymmetries**

39
 40 Given the outlined epistemically complex situation of Ms. Metis, there are impor-
 41 tant epistemic asymmetries between her situation as a potential research partici-
 42 pant and the epistemic situation of the scientists involved in the study. As a first
 43 approximation, it seems intuitive to conceptualize these asymmetries using the
 44 concept of “trust.” Accordingly, participants enter research with an attitude of
 45 trust that research is based on sound principles and that it is more or less safe and
 46 reliable. There is some empirical support showing that participants are greatly
 47 influenced by the information they receive on medical decisionmaking irrespec-
 48 tive of whether that information is evidence based or not.³⁵ This can be interpreted
 49 as indicative of a certain basic reliance or trust.

1 To make a well-informed decision, potential research subjects do not only
2 need to know, among other things, their subjective interests and risk preferences.
3 A well-informed decision requires also some cognitive access by research subjects
4 such as Ms. Metis to what would be in her best interest from a more objective point
5 of view. That is, independent of how Ms. Metis *evaluates* the facts, she must have
6 some understanding of *the facts*. However, to know what is in one's own best inter-
7 est in an epistemically complex situation as outlined may go well beyond one's
8 cognitive agency. This is hampered by external influence such as time pressure,
9 urgency, or affective states such as fear and anxiety or even mild cognitive impair-
10 ments. Moreover, the degree of desperation constitutes a vulnerability of people
11 diagnosed with a severe medical condition for which no effective therapy is yet
12 available. This is "vulnerability" in a nonstigmatizing sense, because it does not
13 devalue the rights of these persons to make their own decisions, given legal capac-
14 ity. Rather it entitles them to the additional right that healthcare providers and
15 physicians abide to even higher degrees to the principles of beneficence and non-
16 maleficence; for example, by setting a "lowered standard of acceptable risk" as
17 some argue.³⁶ But if the abovedescribed account is correct, this should rather
18 include higher standards for the *strength of evidence* that support the probability of
19 "acceptable risk."

20

21 **Epistemic Dependence**

22

23 The simpler a theory is, the better, but sometimes the simplest adequate theory is
24 complex and all simple theories are false. As scientific theories tend to be complex,
25 scientists and clinical researchers themselves acquire specialized epistemic skills
26 and take advantage of a fine-grained professional division of epistemic labor. As
27 such, scientific progress is a collective enterprise. Social practices such as peer
28 review and the critical appraisal of the work of others is essential for the quality
29 assurance and improves the reliability of scientific evidence. In consequence, reli-
30 able evidence is the process of a collective, interdependent, and social process.
31 Similarly, Ms. Metis depends on the cognitive capacities of others; that is, she
32 depends on insights and evidence provided not only by individual researchers
33 working on a relevant research question but also on the scientific community at
34 large that provides the epistemic "infrastructure" and quality control. Ideally,
35 patients can trust that the information provided by researchers is honest, accurate,
36 and reliable and that potential open questions and uncertainties are systematically
37 addressed with due diligence. If this can be granted, it would allow potential can-
38 didates to go beyond what they can know or vindicate by their own cognitive
39 resources independently and self-sufficiently (thesis of epistemic dependence).

40

41 **Epistemic Duties**

42

43 A key task of scientists is to retrieve relevant information on an investigational
44 intervention and to minimize uncertainties for hypothetical benefits and poten-
45 tial hazards. Recent insights from social epistemology in analytic philosophy are
46 valuable for analyzing the duties that arise in such situations. According to reli-
47 abilism, a (true) belief is justified if and only if the belief has the right kind of
48 causal history; that is, if the belief formation is reliable to distinguish between
49 truth and falsity.³⁷

1 For Ms. Metis, this account has several implications. She can be justified in her
2 true beliefs about trial participation, if only the information provided during
3 informed consent process is reliable and is communicated in an intelligible and
4 reliably truth-preserving way; that is, without manipulation, nudging, pressure,
5 deception or any other form of “truth-modulation.” Researchers have developed
6 different tools to increase reliability of belief-formation. The most uncontroversial
7 techniques include the scientific method itself but also cultural, institutional, and
8 organizational structures that increase the reliability of evidence. Beyond that, the
9 most important measure for quality assurance is independent peer review³⁸
10 and critical appraisal of published information in systematic reviews and meta-
11 analyses on the basis of clear reporting criteria.³⁹ The open science movement is a
12 further breakthrough in improving reliability of evidence unraveled by research
13 practices.⁴⁰ In addition, metaresearch is forcefully transforming the scientific
14 enterprise by scrutinizing research methods and “testing empirically their effec-
15 tiveness at producing the most reliable evidence.”⁴¹

16 Because of the high standards of transparency, open science may have some
17 practical inconveniences. The discussion about open science is often framed about
18 individual benefits that override these repercussions⁴² and about which external
19 incentives are best suited to complement intrinsic motivation and support scien-
20 tific virtues. This focus on external incentives and individual benefits is pragmatic
21 but theoretically misdirected. For example, in explorative research such as “first-
22 in-human” DBS research, openness, reproducibility, and transparency are not only
23 virtues of epistemic excellence of outstanding individuals, but rather ordinary
24 epistemic duties that can be directly derived from the practical utility of increasing
25 the reliability of evidence. Given the complexity and interdisciplinarity of the
26 whole clinical translation process from bench to bedside, no individual scientist or
27 team can rationally take full responsibility for minimizing all uncertainties about
28 relevant open question. Nor does it seem feasible to systematically oversee at a
29 given moment in time the potential relevance of published results for future
30 research.

31 Peer-scientists therefore seem to have an epistemic right—if perhaps not a legal
32 one—to comprehensive and unrestricted access to relevant information constitut-
33 ing scientific evidence. This can be seen as a corollary from realizing that pre- and
34 postpublication peer review is partly constitutive of the reliability of the very evi-
35 dence that builds the basis of scientific belief formation

36 It is always conceivable that there is some relevant evidence for a particular
37 research question that investigators should look for as further support or negation
38 of their hypothesis, even if they do not possess any evidence at the moment that
39 there is such evidence.⁴³ The evidence that one expert holds given that particular
40 person’s contingent epistemic situation may not be exhaustive. The remedy is
41 repeated critical thinking by many experts. Ideally, every expert who publishes an
42 article on a hypothesis should benefit from the critical appraisal by its readers,
43 who should be encouraged to raise open questions as part of a continued postpub-
44 lication review. If performed for scientific ends (and not as power game among
45 scientists), competing constant critical appraisal increases the reliability of evi-
46 dence, and may efficiently spot yet-unaddressed uncertainties. An article that
47 holds up to this constant public and critical appraisal may still be empirically dis-
48 proven, but the original evidence provided by the article will stand strongly on
49 theoretical grounds. In this way, the *reliability of the evidence provided* can be

1 increased, although it may not increase the evidence provided by the scientific
2 article per se.

3

4

5 **Problem Summary**

6 High-risk/high-gain research at the frontier of science involves *by definition*
7 unknown probabilities about clinically relevant factors. These relevant factors
8 may sometimes bear on questions about life and death such as brain hemorrhage
9 or serious brain infections. Exploratory research is an approach that severely
10 impedes evidence-based decisionmaking, and involves ineliminable uncertainties
11 about potential harms, potential absences of benefits, and potential futile trial par-
12 ticipation. Therefore, *exploratory research* involving a high magnitude of burden
13 such as brain surgery is hardly justifiable if uncertainty about relevant risks is
14 high. The complexity of estimating the joint probabilities of accumulating uncer-
15 tain factors exacerbates evidence-decisionmaking about trial participation, and
16 may lead to misestimation of risks.

17 In the perspective of a risk taker, it is often the case that potential participants
18 can decide only once whether or not to participate in a high-risk/high gain trial.
19 They have one attempt to hit a home run. In contrast, researchers examining novel
20 investigational interventions; for example, for Alzheimer's disease, may have sev-
21 eral opportunities to engage in high-risk/high-gain research. Trials are expected
22 to frequently fail, and therapeutic interventions are urgently needed. In addition,
23 it is sometimes sufficient to already detect some signal of efficacy only once ("proof
24 of concept"), so that each participant examined is another opportunity to still be
25 successful. It is noteworthy that participants take considerable risks in "first-in-
26 human" studies involving neurosurgery, but—by the definition of "first-in-
27 human"—cannot rationally expect medical benefits from the intervention based
28 on any prior probability that is supported by strong directly relevant empirical
29 evidence. Although the pragmatist stance; that is, to believe in the intervention
30 nonetheless (optimistic bias), is viable for research subjects, researchers ought
31 only to believe and communicate what is *strictly* supported by the evidence to
32 avoid the "uncertainty blindness" of potential research subjects.

33 Strengthening the reliability of evidence or identifying yet-unknown uncertain-
34 ties is a difficult scientific task that is a social endeavor of the scientific community
35 as a whole and that is best reached by constant critical appraisal and reuse of com-
36 prehensively, transparently, and openly published data.

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38

39 **Strength and Limitations**

40 Evidently, a valid argument is only as strong as its premises, and to resist one of its
41 premises easily allows resistance to the whole argument. This conceded, I am con-
42 fident that resisting the premises is not an easy task. Even opponents of the view
43 that scientists are epistemically obliged to transparency, openness, sincerity, and
44 honesty at least agree that scientists should not engage in "wishful speaking."⁴⁴ It
45 is even harder to challenge the view that potential participants of high-risk/high-
46 gain studies are not epistemically dependent on researchers. But if so, the argu-
47 ment is in good standing. The epistemic dependence gives rise to the epistemic
48 obligation of the researcher "to communicate only those claims which are well
49 established"⁴⁵ and to strictly stick with the evidentialist agenda to avoid epistemic

1 perils such as “uncertainty blindness.” However, being an evidentialist entails
2 being a researcher who strives to live up to not only the ideals and virtues of sin-
3 cerity and honesty but also to those of transparency and openness.

4 A clear limitation of the presented account is its scope. I have only defended the
5 existential statement that there are epistemic duties and started to preliminarily
6 characterize the reasons for such potential duties. However, little has been said
7 about the moral force or the enforceability of such epistemic duties.
8
9

10 Conclusion

11 Research participants are entitled to be informed about the *strength of evidence*
12 available to estimate the probabilities of relevant factors potentially influencing
13 the safety or efficacy of an intervention. The mere complexity of intervention
14 ensembles such as DBS for Alzheimer’s disease illustrates how yet-unidentified
15 open questions give rise to a combinatorial explosion of uncertainties. To mitigate
16 this risk demands collective social epistemic practices that no single research team
17 is likely able to provide alone.

18 Open science, transparent and exhaustive data reporting, preregistration, and
19 continued constant critical appraisal via pre- and postpublication peer review seem,
20 therefore, not to be scientific virtues of moral excellence but rather ordinary obliga-
21 tions of the scientific work routine to increase *reliability* and *strength of evidence*.
22

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