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The need for a change in medical research thinking. Eco-systemic research frames are better suited to explore patterned disease behaviors

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Many practicing physicians struggle to properly evaluate clinical research studies – they either simply do not know them, regard the reported findings as ‘truth’ since they were reported in a ‘reputable’ journal and blindly implement these interventions, or they disregard them as having little pragmatic impact or relevance to their daily clinical work. Three aspects for the latter are highlighted: study populations rarely reflect their practice population, the absolute average benefits on specific outcomes in most controlled studies, while statistically significant, are so small that they are pragmatically irrelevant, and overall mortality between the intervention and control groups are unaffected. These observations underscore the need to rethink our research approaches in the clinical context – moving from the predominant reductionist to an eco-systemic research approach will lead to knowledge better suited to clinical decision-making for an individual patient as it takes into account the complex interplay of multi-level variables that impact health outcomes in the real-world setting.

KEYWORDS

systems thinking, research design, philosophy of science, uncertainty, evidence-based medicine, complexity science, philosophy of medicine, complexity thinking

Scientific research traditions

The roots of modern research trace back to the late 17th century with the exploration of the innate (physical) world.

Newton’s research establishing the laws of the innate physical world based on experiments and repeated measurement in the controlled setting of the laboratory. This approach is based on a number of assumptions with limitations in real world applications – firstly, to experiment in the laboratory setting removes all external context that otherwise would impact the experiment (the law of the free fall of an object holds true only in a vacuum); secondly, that one can exactly measure observations (though Gauss showed that repeated measurements always have an error that symmetrically distributes around the mean); and lastly, that repeating the same experiment at a later time in a different setting will result in exactly the same outcome.

About a century later, Goethe and Humboldt demonstrated that Newton's laws of the inanimate world of physics did not apply to the animate world of living beings. To understand and predict their behavior required the simultaneous understanding of their environmental context (1). Furthermore, Pareto observed another important phenomenon of the animate world, namely that it has a consistent distribution pattern that follows an 80/20 split – now known as the Pareto or inverse power law distribution (2). These observations marked the recognition of the interconnectedness and interdependence inherent in biological systems.

Humboldt is regarded as the founder of systems sciences – the sciences of interconnectedness and interdependence within mechanical and biological systems. In general terms, such systems consist of at least two parts where

- the whole cannot be divided into independent parts,
- each part affects the behavior of the other, and
- the way each part affects the behavior of the system depends on what at least one other part is doing (3).

Biological systems have the added characteristic of being adaptive, i.e., the behavior of one part can change the behavior of all other parts. Over time such changes lead to emergent – marginally stable – system states [homeokinetics (4)] which, in the medical context, we associate with particular diseases and disease severities (Figure 1 – top).

Mechanistic vs. eco-systemic research questions

There is a basic difference between physical and biological/social research questions. Physics is concerned with explaining cause-and-effect relationships in the inanimate world whereas biological/social sciences focus on understanding the emergent structural and behavioral phenomena in nature. While physics rightly focuses on researching *mechanisms* through a reductionist research paradigm, biological/social sciences should adopt an eco-systemic approach to understand the ways living beings *behave* and constantly *adapt* at all scales of organization within their changing environments. Biological/social sciences should not only concern themselves with the structure and dynamics of 'biological/social systems', but more importantly with finding *meaning* or *making sense* of those eco-systemic interactions (5, 6).

Medicine is not a science, it is a praxis (7). Clinicians use those scientific results that as good as possible apply to the individual. Given the endless biological/social variability between individuals and their highly variable living environments, they can never deliver perfectly predictable outcomes. Despite these variabilities, our interventions almost always result in one of a number of limited (i.e., not infinite) familiar patterns of outcomes.

The early successes of medicine arose mainly from the insights of reductionist research that explained the 'simple' cause-and-effect mechanisms of then common and life-shortening infectious diseases. However, 21st century medicine mostly struggles with chronic and complex diseases whose successful management demands a systemic understanding of the 'complex' interactions amongst the multiple variables from across the different scales of organization.

Put pragmatically, studying the effect of a defined antibiotic on a defined bacterium, an antihypertensive on blood pressure changes, or

an antidepressant on a change in mood/anxiety scores in the laboratory would rightly be best done using the reductionist cause-and-effect research approach. However, many of these findings produced in the highly controlled laboratory environment are not reproduced in 'real world' clinical trial settings.

Clinically relevant questions necessitate systemic research approaches focused on patient relevant outcomes like:

- Does a new antibiotic work safely in people, and if so, what is the right dose for a particular person?
- Does lowering blood pressure prevent heart attacks or strokes, and if so, how much blood pressure reduction for an individual patient reduces his/her *absolute* risk of an event?
- Which type/combo of therapy/ies is best to recover from trauma or loss, and how does that vary amongst people from different social/ethnic backgrounds?
- Whom does a particular population-based prevention intervention benefit, and what are the issues that make it fail in others?

In the laboratory setting, research typically focuses narrowly on one-to-one relationships in the absence of any other contextual constraints (8). What may work well in the deliberately chosen context-free laboratory setting does not necessarily also work as a clinical intervention in diverse clinical settings. By their very nature clinical events are caused by a multitude of interacting factors. Clinical interventions cause one-to-many interactions simultaneously affecting physiological, environmental as well as sense-making/coping systems. Put succinctly, one-to-many relationships cannot be studied by 'squeezing' them into 'sanitized' one-to-one methodologies (8).

The bottom section of Figure 1 - bottom provides contrasting examples of research questions that either focus on mechanistic cause-and-effect problems or seek to gain insights and understandings of the complex interconnected and interdependent one-to-many cause-and-effect dynamics impacting people's health.

Finding the cause vs. understanding heterogenous outcomes

Researching a cause-and-effect problem like determining whether 'a new antibiotic kills a bacterium *in vitro*' falls within the Newtonian research paradigm. It requires repeating the same experiment to determine the reliability of observations.

In contrast research to understand observable differences, i.e., patterns, related to a particular phenomenon [e.g., blood sugar dynamics in insulin-dependent diabetics (29), or experiencing significant diabetes symptoms despite adequate blood sugar control (30)] requires a different approach. Patterns emerge depending on the interactions and combinations of several contributing variables. Pattern analysis techniques like cluster (31) or network (32) analysis can identify which combinations and interactions lead to each of the observed outcomes of interest, and may guide further research in understanding their 'causal pathways.' Figure 1 contrasts the differences between the two frames, and Figure 2 illustrates how cluster analysis techniques can inform the management of coronary artery disease (33).

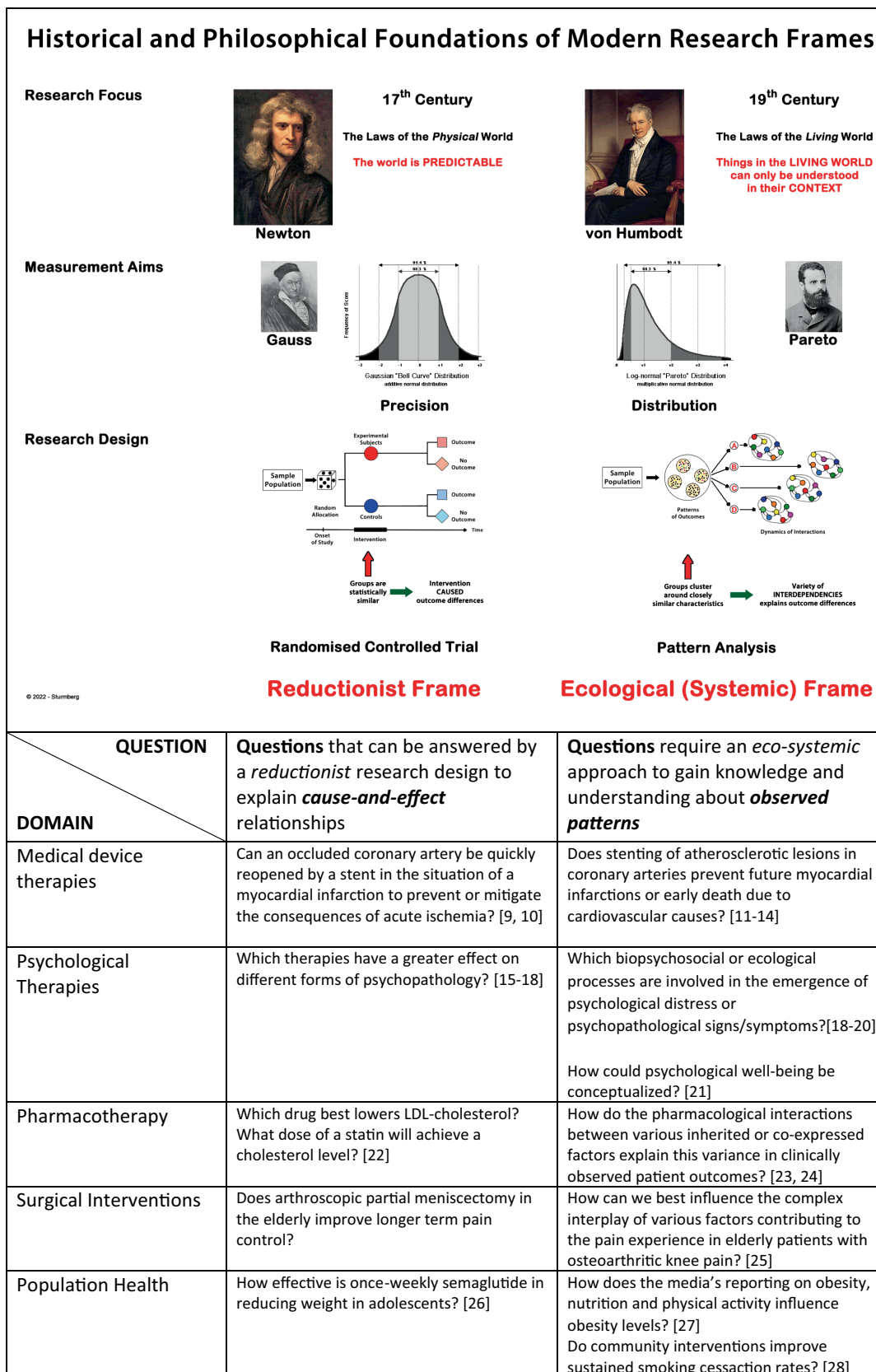


FIGURE 1 Comparison of the reductionist and eco-systemic research frames. Note, that the reductionist approach aims to establish clear and repeatable cause-and-effect relationships, whereas the eco-systemic approach aims to gain insight into the dynamics that result in patterned outcomes. Understanding what "caused" an observed pattern (looking backwards) will allow clinicians to use the "pattern specific" interventions best suited to this patient (looking forward). The table provides research questions that can be best answered within each research frame (the selected references only relate to the nature of research questions rather than the differences in research methodology).

How to measure clinically relevant outcomes?

Clinically meaningful eco-systemic outcome measures can only be direct measures of endpoints such as hospitalizations, mortality or quality-of-life, resulting in so called '*patient-oriented evidence that matters*' (POEM) (34). Clinical research frequently relies on indirect ('*surrogate*') outcome measures in the form of 'biomarkers' like laboratory measures and radiological quantifications that are assumed to indicate 'clinically meaningful' outcomes (34) or a combination of very different clinical ('*composite*') outcomes that tend to overstate benefits (35) (in this context one should consider Goodhart's law¹). However, even though biomarkers may align with pathophysiology of disease, they often fail to reliably predict effects on a clinically meaningful endpoint. For example, clinical trials of lowering the biomarker LDL has had at best tenuous impact on overall survival (36). Even more difficult to define are meaningful outcome measures for psychological/psychotherapeutic interventions – symptom reduction/remission, while common outcome measures, are highly subjective (37) with patients taking what they think and feel most relevant for their lives (38). And finally, the magnitude of an outcome is sensitive to the characteristics of the study population – while an intervention may have only a small benefit at a community level it may result in more people benefiting than the same intervention targeting a high risk cohort (39).

Hence, the *question* we really need to answer is: which patient in which context will *most likely* benefit from an intervention in a subjective and objective way?

Implications

Research, regardless of its methodology and rigor, provides additional data rather than information or knowledge (40). Statistics indicate the probability that – at the population level – these data *correlate* with particular population observations. However, statistical correlation does not equate to *causation*. Statistical correlation can only infer a potential causal relationship with a certain probability, and only if the relationship is based on a strong pathophysiologic rationale (41, 42). Hence, it is the researcher's responsibility to provide critical *contextual interpretation* of new data to justify their integration to existing understandings. As clinicians we must consider the new understandings in relation to their applicability at the individual/population level, but most importantly, in their unique contexts. And finally, research cannot relieve us from the task of making decisions and being responsible for them.

Knowing the '*study patient*'

It is critical to appreciate that there is no 'prototypical' patient who can guide clinical practice. The randomized controlled trial provides crude information about the outcome differences of the '*average patient*' in a study cohort receiving an active versus a placebo intervention. Observed differences, even when statistically significant,

generally only have a very small pragmatic (or absolute) benefit. An intervention that helps 1 in 2 *average* patients (NNT=2) is 50% effective and 50% ineffective, one that helps 1 in 20 (NNT=20) is 5% effective and 95% ineffective, one that helps 1 in 100 (NNT=100) is 1% effective and 99% ineffective, one that helps 1 in 200 (NNT=200) is 0.5% effective and 99.5% ineffective, and so forth (43). Put differently, even so-called 'good medical interventions' are – pragmatically speaking – ineffective for most patients, and the one benefiting is not identifiable from the data. The same applies to harms which often are not expressed in clearly understandable and comparable terms. Of note, in many cases the increase in intensity of an intervention does not improve outcomes but results in increasing harms, e.g., the so called 'J' curve in treating hypertension (44, 45) or the use of non-steroidal anti-inflammatories in acute and chronic pain (46).

Whose interests matter most?

Research, like other societal activities, is shaped by the philosophical (47, 48), political and industry doctrines and vested interests of its time – consider, e.g., Mbeki's stance on HIV (49), or the regulation of embryonic stem cell research (50–52); or industries' influence on research agenda setting (53), financing, conducting and interpreting research (54), or influencing which type of evidence should be prioritized for policy-making (55).

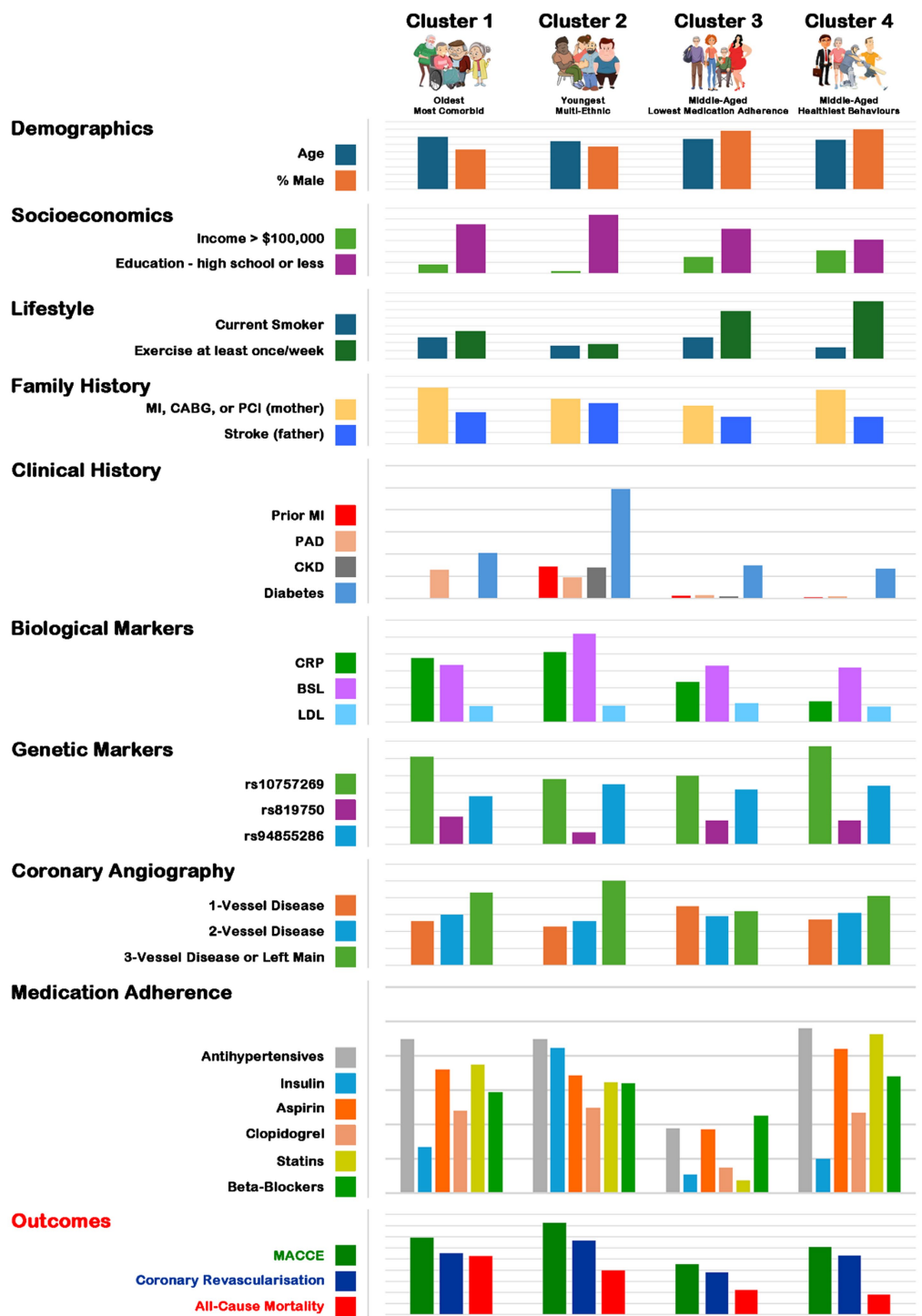
The reductionist understanding that the '*statistically significant*' *dichotomous* outcome difference in a randomized controlled trial implies a '*mechanistic*' cause-and-effect relationship remains widely, but incorrectly, regarded as providing sufficient evidence to promulgate particular pharmaceutical or biomedical interventions to an affected patient population. This misunderstanding suits industry interests well (56). The typical large-scale multi-national industry funded studies only demonstrate small though statistically significant effects, often limited to surrogate or composite outcomes, which are promoted as seemingly benefiting (the misuse of *relative benefit*) a large number of people (euphemistically referred to as 'customers'). The rising trend of accelerated drug approval based on surrogate outcome improvements is of great concern given that more than half of approved drugs do not report confirmatory trial outcomes within the required timeframe causing patient harm and high costs despite uncertain clinical benefit (57, 58).

These observations highlight the significant conflict of commercial versus patient-benefit interest of pharmaceutical/device-maker companies (59) – they have nothing to gain from identifying the small group of patients who will ultimately benefit from a given medication/device (60, 61). Further, applying data from relatively healthy, homogeneous backgrounds to vulnerable patient groups not studied in the trials is fraught. The prevailing focus on biomedical intervention research distracts us to appreciate that greater health improvements are more often achieved by strengthening services that address the social and inequality issues within societies (62, 63).

Can precision-medicine result in better global health?

The precision-medicine movement has recognized the failings of population-based intervention studies and

¹ When a measure becomes a target, it ceases to be a good measure.



Redrawn from Data of 1329 Participants (total of 155 Variables) by:
 Flores AM et al. Unsupervised Learning for Automated Detection of Coronary Artery Disease Subgroups.
 Journal of the American Heart Association. 2021;10(23):e021976. doi:10.1161/JAHA.121.021976

BMI, body mass index; CABG, coronary artery bypass graft; CAD, coronary artery disease; CRP, C-reactive protein; LDL, low-density lipoprotein; MI, myocardial infarction; PAD, peripheral artery disease; PCI, percutaneous coronary intervention

MACCE, composite of myocardial infarction, stroke, coronary and/or peripheral revascularisation

FIGURE 2

Patterns associated with cardiovascular disease outcomes (33). The comparisons should be read across the domains as well as columns. A few notable observations should be highlighted (some are well-known): education and income are associated with better outcomes; a diagnosis of diabetes is associated with greater coronary artery disease burden; CRP levels are high in the oldest multi-morbid and diabetes effected multi-ethnic cluster, while LDL levels are remarkably similar across the 4 clusters; 3-vessel disease is age and co-morbidity burden associated; medication

(Continued)

FIGURE 2 (Continued)

adherence appears to have little impact on disease severity and both, composite and all-cause mortality outcomes. Composite cardiovascular and all-cause mortality outcomes are associated with age and co-morbidities, whereas medication neglect and positive health behaviors have paradoxical associations with composite cardiovascular but no all-cause mortality associations. The difference in coronary revascularization in the latter two clusters may indicate provider bias – non-adherence to medical protocols makes those less deserving, while the health-conscious behavior ones overly deserving of interventions. Redrawn from data of 1329 participants (total of 155 variables) by: Flores et al. (33). BMI, body mass index; CABG, coronary artery bypass graft; CAD, coronary artery disease; CRP, C-reactive protein; LDL, low-density lipoprotein; MI, myocardial infarction; PAD, peripheral artery disease; PCI, percutaneous coronary intervention; MACCE, composite of myocardial infarction, stroke, coronary and/or peripheral revascularization.

aims to discover more specific interventions at the genome/transcriptome/proteome levels. These are expected to be highly predictable to deliver the desired outcome at the patient level (64, 65). Precision-medicine has demonstrated marked improvements in the treatment of certain cancers and improved pharmacotherapy (e.g., warfarin), but has failed to improve interventions and outcomes for common and multimorbid conditions (66, 67).

The promises of precision-medicine may be more wishful thinking than reality (65, 68, 69). Even changes at the physiological level have systemic effects beyond the correction of a specific genomic, transcriptomic or proteomic abnormality. Furthermore, the simplistic understanding that any such ‘precision’ therapy will have a specific target in human biology is fraught and ignores known physiology and pharmacology. Any drug must overcome basic absorption, distribution and metabolism problems even before it comes close to effectively targeting the cell machinery. Additionally, drug effectiveness changes with variability in cell biology, genetic makeup, genomic expression, and change in cell presentation over time (70). Latest at the metabolomic level will we see divergent systemic behavior and less predictable outcomes. Despite these fundamental reservations, an approach to collate the outcomes of individually targeted precision-medicine interventions has the potential to identify community-wide response patterns, an approach that aligns with the eco-systemic research frame.

The way forward

In conclusion, achieving more predictable medical interventions requires a more comprehensive understanding of which systemic variables, and which contexts, lead to the variety of our observable outcome patterns. Recent systems-focused research has demonstrated improvements in diabetes management (71), the drug treatment of hypertension (72), understanding the treatment of depression (73) and the treatment of brain tumors (74) but is, at this stage, a notable exception in clinical research. More systems-focused studies would significantly contribute to the knowledge required to define which outcome pattern a patient – and especially those with multiple morbidities – most likely will belong to. Understanding the underlying bio-medical, social, emotional and interpersonal features (75) underpinning outcome patterns would then enable us to offer the *most likely* treatment to remedy the issue of concern.

Learning to cope with uncertainty

One of the challenges to achieving this goal is our psychological need for certainty in clinical decision-making under always uncertain circumstances. The mental frame of evidence-based medicine as outlined by Sackett et al. remains widely seen as the best possible solution – “*integrating clinical expertise with the best available external evidence from systematic [meaning clinically relevant] research*” (76) in clinical decision-making for this particular patient. However, the best available evidence remains insufficient, which is something that patients and doctors alike should be painfully aware of, but neither are comfortable to acknowledge in a fully open and transparent way. Unwittingly, they collude, in Richard Smith’s words, in a “*bogus contract*” (77).

Medical education, industry and the media all reinforce the socialization of medicine’s unquestionable grandeur. Collectively we rid ourselves of the discomforts of uncertainty by using the mental trick of “*causal inference’ as a tool ... to determine a cause by observing an effect*” (78). We fail to see the circularity in the argument – an ‘*observed effect*’ suddenly is the new cause for ‘*another observed effect*’ and so forth (79). Having, what seems to be, a rational argument allows us to confidently justify the widespread use of therapeutic approaches of limited to minimal effectiveness.

Embracing the inherent complexities

While this discourse outlines the philosophical and methodological underpinnings of medical research thinking, it calls for pragmatically considering the inherent complexities facing medical research and practice. From a science perspective, studying biological/social systems with their nonlinear distribution patterns requires different methodological research approaches. From a professional perspective, medical interventions are system-wide interventions, and their impacts always need to be considered across the molecular to environmental scales. From a practitioner perspective, even the most appropriate and most diligent research trial will always only give an *approximate* answer, and it ultimately at best reduces some degree of a clinician’s uncertainty when having to make decisions in the context of the patient in front of them (80). And from a societal perspective, it challenges the usefulness of medical guidelines as much as the listing and/or public reimbursing of many drugs and medical interventions, like the suppression of ventricular ectopic beats with flecainide (81), the mortality benefits of colorectal cancer screening (82, 83), the

effectiveness of molnupiravir on hospitalization or death (84), or knee arthroscopy for degenerative osteoarthritis (85).

Concluding thoughts

In summary, the reductionist medical research of the late 19th/early 20th century undoubtedly has led to great benefits in understanding and treating the predominant infectious diseases of the time. However, it failed to achieve the same benefits in relation to the now predominant chronic and multimorbid conditions affecting our patients. These problems are systemic in nature, i.e., they are the result of interconnected and interdependent activities spanning from the gene to the societal level. From a pragmatic perspective, we need to firstly shift our way of thinking toward an eco-systemic frame, and secondly, need to further develop the as yet embryonic eco-systemic research tools to find those solutions that allow us to offer the *most likely* beneficial approaches to each of our patients. And lastly, there is an urgent need to re-orientate our undergraduate medical courses to develop critical analytic thinking, and to teach our post graduate specialty trainees a wide range of research methodologies beyond the RCT.

Data availability statement

The original contributions presented in the study are included in the article/supplementary material, further inquiries can be directed to the corresponding author/s.

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