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Introduction

We live in a time of great change. The 2020 pandemic adds to and amplifies three socio-technical and techno-economic processes, which are at the basis of three joint crises: biological-environmental (climate change), techno-productive, and economic-financial. It seems that the whole world is approaching what scholars of many disciplines define situations in which there are high risks of “unwanted collapse” (Scheffer et al. 2012: 344). An unwanted collapse takes the form of a tipping point, that is, a catastrophic bifurcation point, “where a minor trigger can invoke a self-propagating shift to a contrasting state” (ibidem: 344). The three joint crises behind the potential global tipping point occur after decades of evolutionary acceleration induced by an array of factors and forces leading towards a “hyper-connected” world; this, in turn, is characterized by the emergence of a multiplex of self-organized local and global interaction structures and processes within and between different domains.

The claim that the socio-technical and economic system is on the verge of a catastrophe is recurring, and often appear at particularly relevant historical nodes – for example in phases of transition between established and upcoming techno-economic paradigms. To put it with Gramsci: “the crisis consists precisely in the fact that the old is dying and the new cannot be born; in this interregnum a great variety of morbid symptoms appear” (Gramsci and Hoare, 1971). This explains the cyclicity (and, thus, the recurring fads) of the theories of cycles, such as that of Regulation, of World-Systems, or of Long Waves (Silverberg, 2003). However, not all catastrophes – in the sense of catastrophe theory – are alike. In this paper, we make the case for the uniqueness of the current unwanted collapse to come, and discuss how a view taking into account the transformations now in the making should inform decision-making.

In order to do that, we present a map of the deep and extended changes that are currently unfolding and shaping the hyper-connected world we live in. This map is described by a collection of fundamental coordinates: building blocks, concepts, and principles that we single out in details while we also highlight their interconnections. The core notion is that of the unprecedented cyber-physical universe taking shape – an informational and physical landscape drawing the boundaries for actions and transformations to unfold. Our contribution lies in the unique combination of several literature strands, which we use to stress the convergence of different and often non-proximate scientific domains around similar perspectives and problems. The snapshot analysis we propose is necessarily systemic, complex, non-linear and recursive, as it mirrors the manner in which the landscape on which

actors operate is characterized. We use our map to derive implications for decision-making; in particular, we suggest that a shift to “adaptive strategic thinking” is required for actors to survive and succeed in the novel landscape.

1. Coordinates for the current era: the “Earth-System” and a cyber-physical universe

We ground our analysis on a set of (improperly speaking) axioms regarding the nature of socio-techno-economic systems. These basic statements define the paradigm of analysis from which our insights descend. They are: 1) the wave like properties of the evolution of complex adaptive systems. 2) The adoption of a general definition of technology as “any intentional extension of a natural process, that is, of processing of matter, energy, and information that characterize all living systems” (Beniger, 1986: 9). 3) The proposition that “A society cannot develop unless an adequate infrastructure for the movement and processing of matter, energy, and information already exists” (Beniger, 1986: 184). 4) Material and immaterial infrastructures are evolving multilayered networks of structured knowledge. 5) Sequences of socio-technical landscapes are complex and interrelated processes evolving towards asymptotic stationary equilibria, while are every now and then interrupted by distributed discontinuities. 6) Deep and extended changes occur when founding rules are changed and their effects propagate even after a long time.

In the introduction, we mentioned the three distributed discontinuities (crises) that are impacting and transforming – potentially in an abrupt, unwanted-collapse-manner – the socio-technical landscape, as indicated in point 5) above. In this Section, we outline the contours of the new landscape in the making and the dynamic forces shaping it. Our idea is that complex developments, approximated in 1) and 4) are producing 6); we outline these complex developments below.

The exponential increase of computational power and storage capacity, the pervasiveness of information processing devices (*Ubiquitous computing*), the creation of software systems able to process an increasing amount of information flows from all over the world (*Ubiquitous connectivity*), and advances in digital technologies and Artificial Intelligence (AI) have been key drivers of a generative process of intermingling networks at multiple scale and across traditionally different socio-economic activities. This generative process has definitively decoupled the locus of value generation and that of information production and processing, which now is ubiquitously distributed. Thanks to the consequent triggering of cross-scale positive feedback among these dynamics, thereby feeding

scenario of multiple complex systems and nested sub-systems, which interact and evolve within the “Earth-System” (Lenton et al., 2008), the outcome of the transformation is a widespread uncertainty and a looming instability. However, a set of emergent patterns is also taking shape. These are: 1) the development of self-organizing processes, able to manage hyper-scale infrastructures, which have been essential drivers of the formation of hyper-structures (Baas, 2013; 2016). 2) The techno-scientific advancements allowed representing (codifying) real processes and outputs from the nanoscale to the ordinary and global scale – especially as every phenomenon can be read through the lenses of information (O’Connor et al., 2019). Each codified “object” in the natural world becomes a source of fine-grained, real time digital data and can be paired with its own “digital twin”, a full description of the object embedded in software systems, which can be used to simulate real-time changes and interventions. This kind of theoretically complete 1:1 map from the subatomic world to whatever level deemed appropriate for designing processes and outputs leads, in essence, to perpetual self-production. Indeed, interaction, feedback and exchange of information created what Zitttrain (2006) has called “generative space” of ideas and knowledge. 3) The closed world of Newtonian theory, to paraphrase Koyré (1957), is over. Human beings live now in an open-ended universe (Kauffman, 1996; 2009), which is continuously expanding and evolving. 4) As a result, we experience a sort of accelerated expansion of the digital universe, parallel and tightly linked to real processes and their dynamics. We can define this complex and dynamic intermingling the *cyber-physical universe*, within which real and digital processes interact and influence each other to the point that sometimes it becomes impossible to distinguish real from virtual. 5) The openness of this cyber-physical universe implies that the Newtonian mechanistic clockwise “in which big problems can be broken down into smaller ones, analyzed, and solved by rational deduction” (Plsek and Greenhalgh, 2001: 625) is no longer working. The “machine” metaphor is out of date, and completely not appropriate to understand what is happening within the Earth-System (Steffen et al., 2007), where the standard model based on linear cause-effect relations does not work. Indeed, globalization of processes, within which goal-oriented interactors (individuals, collective entities) pursue their goal(s), give rise to interlocking relationships, with relational topologies emerging from exploratory activities performed in different techno-scientific search spaces. The cyber-physical universe becomes the world of non-linearities, because agents populating it evolve on the basis of exchanging information, constructing and modifying systems of beliefs, cognitive procedures, mental models and system of rules – all endogenously shaped by the topology and nature of multi-level and multi-domains interactions. These non-linear and systemic dynamics of cross-influences has triggered an exponential acceleration of change of the Earth-System on many levels.

Taking stock, profound techno-economic transformations and their co-evolution extended to the whole Earth-System are producing a novel, unprecedented landscape on which actors operate – the cyber-physical universe. This landscape is characterized by multi-level complexity, non-linearities, and is made coherent by the pervasiveness of its informational nature. Given that, and paraphrasing David Deutsch (1998), we advance the following statement: *The entire Planet has become a techno-social system, where Information Technologies constitute the “Fabric of Reality”*.³

Such an unprecedented configuration of reality necessarily shapes the set of opportunities and challenges actors face. At this point in the discussion, three issues deserve to be addressed: 1) What made these advances in techno-science possible? 2) Given the fundamental transformations we outlined, how has changed the decision-making landscape around us? 3) On which mental models (paradigms) should decision-making processes be based, now that we are immersed into the cyber-physical universe?

2. The trajectory over centuries towards a cyber-physical universe. What made these advances in techno-science possible?

In this Section, we outline three fundamental steps in the evolution of human attempts to represent the world around us. Cumulatively, these steps have set the stage for the cyber-physical universe to take shape.

1st step: the discovery that the written language of the world can be binary

Philosophers have always questioned the nature of mathematics and geometry, as well as their relationship. A watershed event certainly was the publication of Galileo's *The Assayer*, where the scientist states that the universe is an all-encompassing book written in mathematical language. For

was born the von Neumann architecture, which is the embedding into hardware of the sequential (Turing) model of computation (Prytkova and Vannuccini, 2020), still the prevalent architecture on which today's computers are based.

At the end of these three major steps, Leibniz's dream of creating a "*characteristica universalis*", that is, a symbolic system capable of representing with the binary system, beyond the syntactic differences existing between the various languages, human thought and all the fundamental concepts and real processes, seems to have come true. In reality, the developments we described set us on a path leading well beyond Leibniz's dream. The binary system and the von Neumann architecture led the way to information technologies, which have been enhanced in the last few decades to the point of becoming what we have called the fabric of reality, a fundamental infrastructure, which in turn interacts and is in a superposition with physical processes to form a global whole – the cyber-physical universe. In the cyber-physical universe, countless sources of information and novelties are continuously generating unexpected impulses: individual and societal needs widespread at the international level; need for strategic resources, such as food, energy, water, or Rare Earth Elements (Balaran, 2019); techno-scientific advances; competitive pressures between companies and countries. Given the nature of the new fabric of reality, it is necessary to rethink the modalities and mechanisms of decision-making processes, as they intervene in (and im

By ontology of agents, we mean a conceptual space that they themselves construct and define according to their ability to frame processes and events, representing the real world and the entities that populate it. In the present era, the ontological space must be defined in relation to new components, which we introduce in the next sections, in the light of the unfolding interlocking relationships among nested networks and processes at the global level.

3.2. Cyber-physical systems within a cyber-physical universe

Given the continuous and enormous expansion of the *info-sphere* (Floridi, 2014; Handy, 2015) and of what Brian Arthur calls *The Second Economy* (Arthur, 2011)⁵, real activities unceasingly generate signals and information, thus giving rise in human minds to an ontological space teeming with multidirectional interconnections between cyber-physical systems, which “integrate”. We consider CPS the *crucial agents* (the relevant unit of analysis) populating the new ontological space, as they are “composed of physical subsystems together with computing and networking.” (Lee and Seshia, 2017: 12) and “integrate physical dynamics and computational systems” (ivi: 77).⁶

3.3. Humans and their External Memory Field

While CPS are the fundamental agents of the new ontological space, another fundamental component is worth examining: what Merlin Donald calls the External Memory Field (or EXMF). In Donald’s words, “the EXMF usually consists of a temporary array of visual symbols immediately available to the user. The symbols are durable and may be arranged and modified in various ways, to enable reflection and further visual processing” (Donald, 1991). In analyzing the evolution of the human mind and cognition, Donald distinguishes three transitions in the representational systems created by the brain during evolution. The adaptive emergence of the third one is the extension of “visuocognitive operations into, and becoming a part of, an *external symbolic system*” (italics added; Donald, 1991: 274). Starting from the invention of written language, a lot of graphic and visual tools have been created through interactions among people and more in general with the operation environment. This evolution has been the result of the attempt to bridge an ever-renewed gap between acquired knowledge and at the same time the need for new knowledge to solve problems. Indeed, the “symbolic use of graphic devices” has been enriched over the centuries through different forms of expression (artistic, technical, scientific, and so on). This unfolding has come about until the turning

⁵ “All digital processes are conversing, executing, interacting, updating, transforming, triggering changes... It is vast, silent, unseen, autonomous, parallel, self-configuring, self-organizing, self-architecting, self-healing.... [like] an aspen root system” (Arthur, 2011).

⁶ “Cyber-Physical Systems (CPS) are integrations of computation with physical processes. Embedded computers and networks monitor and control the physical processes, usually with feedback loops where physical processes affect computations and vice versa”. (Lee, 2008: 363).

point (we add) of the binary system proposed by Leibniz, who provided an extremely powerful impetus for the development of an essential cognitive workspace, thanks to a symbolic system inherently tending to represent the world in its entirety (universality), starting from basic principles and building on them rules and systems of rules.⁷

Since CPS combine computation, communication, and physical dynamics, while the EXMF has become an expanding universe of both organized information flows and chaotic information particles, it is not surprising that human mental frames have been striving to pursuing ever-greater computational power and ever more sophisticated representational systems. Indeed, cumulative feedback loops between the application of information-processing devices to the production of new information have powered a sort of arms race between knowledge-accelerated growth and the tools to master it. The binary system was therefore a key driver in feeding the continuous expansion of the info-sphere and consequently of making information technologies the fabric of generativity, as defined by Zittrain (2008: 70): “*Generativity is a system’s capacity to produce unanticipated change through unfiltered contributions from broad and varied audiences*”. Generative systems occur when information flows, possibly coming from countless sources, self-organize based on the congruence between shared interests, values, paradigms and worldviews, or “simply” since the agents share compatible research guidelines and objectives. The novelty of our age is that, thanks to information technologies, generative systems *are drivers and result* of global interconnections, so that they show particular features: 1) *scalability*, due to ubiquitous computing and connectivity. 2) *Adaptability*, as physical architecture and software systems unceasingly evolve, in this way allowing more and more information be created and/or processed. 3) *Progressive blurring of boundaries* between material and immaterial processes, thanks to their integration realized by the globally spreading of CPS.

other outputs, which are yet unknown when they are invented. Those materials do not exist in nature – as far as we know – and are created by engineering them at the atomic and sub-atomic level, or at the nano-scale, and then built-up to the scale of everyday life, in what is labelled multilevel materials

advanced robotics, Internet of Things), new materials (Bio- and nanomaterials, supermaterials), new techno-scientific processes (data-driven production cycles, synthetic biology, post-genomics, data-driven scientific discovery, and applications of artificial intelligence systems). The outcome of these feedback loops is the dissemination of knowledge-intensive processes and outputs, where interdependencies, complementarities, cognitive and operational conflicts, systemic integration become essential dynamic properties. In this scenario, it is not surprising that the boundaries between firms are no more crisp, but rather “fuzzy and blurred”, and they tend to be conceived in terms of innovative eco-systems (Paulus-Rohmer et al., 2016). In fact, we can claim that the study of companies’ decision-making in the new landscape will miss the mark if it will continue to focus on the firm as unit of analysis. The correct unit of analysis is rather *the bundle of organized process that harness distributed information flows from the concert of sources we singled out* in this paper. This means that the mechanism of formation of firms’ boundaries cannot be fully proxied by the classic “make or buy” trade-off or by simple transaction costs arguments. Firms, as micro-organisms in symbiotic (though not always healthy) relation with the Earth-System and immersed in the cyber-physical universe, are subject to continuous structural re-modulations, given the complex, ever-changing pressures and opportunities at multiple levels.

At the level of the economic and productive sequences, the variable sets of phases and operational tasks (Baldwin, 2012; Yonatamy, 2017) can be modeled with computational tools on the basis of a systemic, multi-scale and integrated perspective. This frequently includes the design of the structural properties of processes, the charact

collected data and information. It is necessary that they perform dynamic functions, in order to support the management of material and immaterial flows, as well as the creative interpretation of increasing information flows. However, the ever-expanding cyber-physical universe on a global scale incessantly generates sets of problems that scientists, experts and researchers from various disciplines strive to solve. A fundamental problem is that of making intelligible the growing mass of data and information, transforming them into useful knowledge to face perennially emerging economic needs. A logical implication of all this is that there is always the need to overcome the gap between the computational capacity of agents (individual and collec

process information based on “declarative knowledge bases”. In this case, the knowledge relating to the domain of a problem is represented through “declarative sentences” and it is processed through first-order logic. While classic AI analyzes well-defined problems using deductive logic rules, the second approach is the sub-symbolic paradigm, explicitly inspired by the biological neural systems of the brain. Starting with the seminal book by Rumelhart and McClelland (1986), neural computing, also known as connectionist approach, models processor-node networks without explicitly representing knowledge through symbols. All (artificial) neural networks are directed graphs processing input into output having defined a certain activation functions for the nodes of the graph (Prytkova and Vannuccini, 2020); modern neural networks extend such topology to encompass multi-layered directed graphs and more modular and hierarchical structures (e.g. “capsules” as in Sabour et al., 2017). The approach (from the initial experiments with perceptrons to current bio-inspired AI) tries to simulate the individual and collective dynamics (rules of activation and propagation of information) of the neural networks that are activated in the brain.

After the first successes, the 90s saw the latest among the cyclical “winters” of AI, because even the connectionist models (initially with only three layers of neurons) seemed to show limits in emulating cognitive functions such as language processing, perception, memory. The consequence was loss of interest, reduction of investments in the research trajectory, and stasis in the creation of new, more sophisticated computational models. The connectionist approach gained new life in the early 2000s, when a group of researchers from the University of Toronto, led by Geoffrey Hinton, introduced the Deep Learning technique. In short, Deep Learning applies the backpropagation algorithm based on gradient descent to update nodes weighting to a new organizational model of the artificial neural networks, made up of many layers (and thus “deep”), with groups of modules in each of them and transversal connections in an impressive numbers (billions). Deep Neural Networks (DNN) models are showing remarkable performances in the recognition of spoken and written texts, images, simple phonemes, reconstructing complex representations from simple and scattered typological details or categories. For example, a type of DNN, convolutional neural networks, uses the operation of convolution to extract feature from complex data input (e.g. images as grids of pixels), layering up these features from the most essential (corners, contours) to more articulated ones (full objects).

The success of Deep Learning in combination with artificial neural networks and the universe of new techniques and refinements developed in the last decade (for example parallel advances in the technique of reinforcement learning) could not have been achieved without impressive advancements in computing power and in the data availability (see Prytkova and Vannuccini, 2020). Increasing

computational power and data availability are the byproduct of the unfolding dynamics that lead to ubiquitous computing and connectivity – the generalized digitization of physical objects and processes that is at the core of the cyber-physical universe.

The last twenty years have witnessed an impetuous development of computerized systems and artificial agents capable of performing tasks and functions that normally require human intelligence. New methods and procedures with genetic algorithms turned out in the planning and control of optimization processes, while models based on neural networks have gradually assumed an increasingly important role in the recognition and processing of natural language and in artificial vision. Several scholars have developed Bayesian models of computational processing, which combine structured knowledge representations with statistical inferential machines. Hierarchical Bayesian models have made it possible to discover “correct structural forms of many real-world domains” (Tenenbaum et al., 2011), as well as causal relationships and analogical transfers of knowledge in different domains. At the origin of these approaches are the contributions of Pearl (1988), Muggleton and De Raedt (1994), and Richardson and Domingos (2006): Pearl has developed Bayesian probabilistic models of causal relationships; Muggleton and De Raedt have contributed significantly to the inductive logic programming trend, which aims to create artificial systems capable of learning autonomously, through what is called statistical relational learning (De Raedt and Kersting, 2017); Richardson and Domingos introduced Markov Logic Networks, which consist of

The changes analyzed in the previous pages imply the need for a paradigm shift for decision making, in terms of general principles and operational criteria. Regarding the principles, the emerging problems around the world require designing systems capable of withstanding temporary and structural shocks through the acquisition of resilience and robustness. Following Folke et al. (2004: 558), we define resilience as “The capacity of a system to absorb disturbance and reorganize while undergoing change so as to retain essentially the same function, structure, identity, and feedbacks”. Instead, robustness of a system indicates the decisional and structural flexibility suitable for absorbing in the long-term changes induced by fluctuating environments (NRC, 2007: 33-34). Studies on this subject show two general principles favor both properties: redundancy and modularity. Redundancy (“the property of one component to perform another’s function”, NRC, 2007: 49) can avoid catastrophic effects resulting from the loss of specific components, such as to generate cascading effects in the event of systemic interdependencies. Modularity means “compartmentalization, or the decomposition of a system into discrete units, into subsets of entities with high-frequency interactions between them and low-frequency interactions between subsets” (Simon, 1962). Modularity confers strength, because it reduces the possibility of spreading negative impulses, similarly to the “social distancing” the world is experiencing in the present era.

From these strategic principles derive no less important operational criteria to be followed in a

Through continuous approximations, we focused first on the emergence of the Earth-System as the common playground or workspace in which processes of different nature take place. Then, we singled out the historical steps that drove humanity towards the cyber-physical universe, and described some identifiable patterns occurring in the current landscape. From there, we unbundled the complexity that characterizes the intertwined forces at work in this context in order to shed light on which opportunities and challenges actors face in the novel ontological space opened by the establishment of the cyber-physical universe. In particular, we described transformations in the mode of conducting production activities, in the boundaries of the firm, and in decision making-processes. The latter, in particular, is subject to a proper paradigm shift, due to advances in AI that can help and augment decision-making. The dynamic matching between progresses in autonomous decision making system and restless mutations in the search-problem-solution space – both byproducts of the rapid evolution of the cyber-physical universe – requires the adoption of principles and operational criteria that are less inspired by the command-and-control approach and more fit to capture the non-linearities of the novel landscape. We suggest that a frame based on adaptive strategic thinking would be the most appropriate for actors to continue thriving in the current era of deep and extended changes.

References

- Arthur B., 2011, The Second Economy, McKinsey Quarterly, October.
- Baas, N.A., 2016. On higher structures. *International Journal of General Systems*, 45(6), pp.747-762.
- Baas, N.A., 2013. On structure and organization: an organizing principle. *International Journal of General Systems*, 42(2), pp.170-196.
- Balaram, V., 2019. Rare earth elements: A review of applications, occurrence, exploration, analysis, recycling, and environmental impact. *Geoscience Frontiers*, 10(4), pp.1285-1303.
- Beniger J.R., 1986, The Control Revolution. Technological and Economic Origins of the Information Society. Harvard University Press.
- Castelvecchi D., (2017), “The shape of things to come”, *Nature*, Vol 547, 20-7, 272-274.
- Davis M., 2000, The Universal Computer, Norton & Company.
- De Raedt L., Kersting K., (2017), “Statistical Relational Learning”, in: Sammut C., Webb G.I. (Eds), *Encyclopedia of Machine Learning and Data Mining*.
- Deutsch, D., 1998. *The fabric of reality*. Penguin UK.
- Domingos, P. and Lowd, D., 2019. Unifying logical and statistical AI with Markov logic. *Communications of the ACM*, 62(7), pp.74-83.
- Floridi L., 2014, The 4th Revolution. How the Infosphere is reshaping the humanity, Oxford University Press.
- Folke et al., 2004, Regime Shifts, Resilience, and Biodiversity in Ecosystem Management, *Annu. Rev. Ecol. Evol. Syst.*, 35: 557–81.
- Gibney, E., 2015. 2D OR NOT 2D. *Nature*, 522(7556), p.274.
- Glaser A., 1971, History-of-binary and other non decimal numeration, Anton Glaser Copyright.
- Gramsci, A. and Hoare, Q., 1971. *Selections from the prison notebooks*. London: Lawrence and Wishart.
- Gödel K. 1931, reprinted in 1992, On Formally Undecidable Propositions of Principia Mathematica and Related Systems, Dover Publications.
- Handy, C., 2015. The Seductions of the Infosphere. *Harvard Business Review*.
- Hertzman A., Fleet D., 2012, Machine Learning and Data Mining. Lecture Notes, CSC 411/D11, Computer Science Department University of Toronto.
- ICME (Integrated Computational Materials Engineering), Ghosh S., Woodward C., Przybyla C. (Eds.) 2020, *Advancing Computational and Experimental Methods*, Springer.
- Kauffman, S., 1996. *At home in the universe: The search for the laws of self-organization and complexity*. Oxford university press.

- Kauffman, S., 2009. Towards a post reductionist science: the Open Universe. *arXiv preprint arXiv:0907.2492*.
- Koyré A., 1957, From the Closed World to the Infinite Universe, Joh Hopkins Press, Baltimore.
- Lee E.A. (2008), “Cyber Physical Systems: Design Challenges”, 11th IEEE Symposium on Object Oriented Real-Time Distributed Computing (ISORC), 363-369
- Lee E.A, Seshia S.A., 2017, Introduction to Embedded Systems. A Cyber-Physical Approach, MIT Press.
- Lenton T.A., Tipping elements in the Earth’s climate system, PNAS, February 12, 2008 vol. 105. no. 6: 1786–1793.
- Marcus, G., 2020. The next decade in ai: four steps towards robust artificial intelligence. arXiv preprint arXiv:2002.06177.
- McDowell, D.L. and Olson, G.B., 2008. Concurrent design of hierarchical materials and structures. In *Scientific modeling and simulations* (pp. 207-240). Springer, Dordrecht.
- Muggleton S., De Raedt L., (1994), “Inductive Logic Pogramming: theory and nethods, *Journal of Logic Programming*, 19 (20), 629-679.
- National Research Council NRC, 2007, *New Directions for Understanding Systemic Risk: A Report on a Conference Cosponsored by the Federal Reserve Bank of New York and the National Academy of Sciences*. Washington, DC.
- O'Connor, M.I., Pennell, M., Altermatt, F., Matthews, B., Melian, C. and Gonzalez, A., 2019. Principles of ecology revisited: Integrating information and ecological theories for a more unified science. *Frontiers in Ecology and Evolution*, 7, p.219.
- Olson G.N., (1997), “Computational Design of Hierarchically Structured Materials”, *Science*, 277, 1237-1242
- Pearl J., (1988), Probabilistic Reasoning in Intelligent Systems, Morgan Kaufmann, San Francisco, CA
- Plsek, P.E. and Greenhalgh, T., 2001. The challenge of complexity in health care. *Bmj*, 323(7313), pp.625-628.
- Prytkova, E. and Vannuccini, S., 2020. *On the Basis of Brain: Neural-Network-Inspired Change in General Purpose Chips*. SPRU-Science Policy Research Unit, University of Sussex Business School.
- Reeves R-, Levin S., 2017, Think Biologically. Messy Management for a Complex world, Boston Consulting Group Henderson Institute, July 17.

- Paulus-Rohmer, D., Schatton, H. and Bauernhansl, T., 2016. Ecosystems, strategy and business models in the age of digitization-how the manufacturing industry is going to change its logic. *Procedia CIRP*, 57, pp.8-13.
- Richardson, M. and Domingos, P., 2006. Markov logic networks. *Machine learning*, 62(1-2), pp.107-136.
- Rudas I.J, et al., 2010, Computational Intelligence in Engineering, Springer.
- Rumelhart D., McClelland J., 1986, *Parallel Distributed Processing*, vol. I, Cambridge, MIT Press MA
- Sabour, S., Frosst, N. and Hinton, G.E., 2017. Dynamic routing between capsules. In *Advances in neural information processing systems* (pp. 3856-3866).
- Scheffer M.et al., 2012, *Anticipating Critical Transitions*, Science, vol. 338, 19, 344-348.
- Schneier B., We Need to Embrace Inefficiency to Save Our Economy, Belfer Center, Harvard Kennedy School, June 20.
- Silverberg, G., 2003. *Long Waves: Conceptual, Empirical and Modelling Issues* (No. 015). Maastricht University, Maastricht Economic Research Institute on Innovation and Technology (MERIT).
- Simon H. A., 1956, “Rational choice and the structure of the environment”, *Psychological Review*, 63, 129–138.
- Simon H.A., 1962, The Architecture of Complexity, Proceedings of the American Philosophical Society, Vol. 106, No. 6, 467-482.
- Simon, H. A., 1991, Models of my life. Cambridge, Massachusetts, The MIT Press MA.
- Steffen W. et al., 2018, *Trajectories of the Earth System in the Anthropocene*, PNAS, August 14 ,vol. 115, no. 33, 8252–8259.
- Tenenbaum J.B., Kemp C., Griffiths T.L., Goodman N.G., (2011), “How to Grow a Mind: Statistics, Structure, and Abstraction”, *Science*, Vol. 331, Issue 6022, 1279-1285.
- Yonatany, M., 2017. Platforms, ecosystems, and the internationalization of highly digitized organizations. *Journal of Organization Design*, 6(1), p.2.

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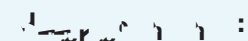
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