



1 Scientific Discovery Reloaded

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5 Abstract

6 The way scientific discovery has been conceptualized has changed drastically in the last few decades: its relation to logic,
7 inference, methods, and evolution has been deeply *reloaded*. The ‘philosophical matrix’ moulded by logical empiricism and
8 analytical tradition has been challenged by the ‘friends of discovery’, who opened up the way to a rational investigation of
9 discovery. This has produced not only new *theories* of discovery (like the deductive, cognitive, and evolutionary), but also
10 new ways of *practicing* it in a rational and more systematic way. Ampliative rules, methods, heuristic procedures and even a
11 logic of discovery have been investigated, extracted, reconstructed and refined. The outcome is a ‘scientific discovery revo-
12 lution’: not only a new way of looking at discovery, but also a construction of tools that can guide us to discover something
13 new. This is a very important contribution of philosophy of science to science, as it puts the former in a position not only to
14 interpret what scientists do, but also to provide and improve tools that they can employ in their activity.

15 **Keywords** Logic · Discovery · Heuristics · Reasoning · Psychology · Algorithm

16 1 Scientific Discovery: The Matrix

17 A long-standing and influential tradition has shaped the way
18 scientific discovery has been accounted for. It has been put
19 forward in particular by logical empiricism, mathemati-
20 cal logic and the analytical tradition in philosophy, which
21 moulded the matrix, that is, the origin and the conceptual
22 framework, of the received theory of scientific discovery.
23 It simply maintains that there is no way of accounting for
24 scientific discovery in logical or even rational terms: “there
25 is no such thing as a logical method of having new ideas, or
26 a logical reconstruction of this process. [...] every discovery
27 contains an ‘irrational element,’ or ‘a creative intuition,’ in
28 Bergson’s sense” (Popper 1961, 32). This idea, which opens
29 up the way to the ‘psychology of discovery’, breaks down
30 into two approaches.

31 The first argues that scientific discovery is a *black box*:
32 the final hypothesis is the only thing that we can see, and
33 this is the outcome of a subjective, idiosyncratic, completely
34 personal process and, as such, it cannot be reconstructed by
35 rational means: “there are extraordinary aspects of the per-
36 son who is able to produce significant new works” (Weisberg
37

2006, xii), and they are essential for discovery. Genius (see
e.g. Murray 1989), illumination, ‘faculties’ such as intui-
tion, insight, or ‘divergent thinking’, are common notions
employed to support this thesis. Even if this line of argu-
ment ends up with a *obscurum per obscurius*, in principle
that is not a problem: the whole process of discovery does
not matter so much since what philosophy and science can
reasonably do is evaluate a hypothesis only *after* it has been
generated—since generation and justification are logically,
temporally and in methods separated. This line of argument
is maintained by famous philosophers like Popper (1961)
and Laudan (1977, 1981) as well as scientists like Einstein
(1958) and Frege (1960).

The second approach argues that even if discovery is a
matter of psychology, we can understand it and account for
several properties of it. This view has been popularized by
Poincaré (1908), and refined and extended by Wallas (1926),
Hadamard (1958) and Simonton (1988). Wallas made
explicit the stages of the process of discovery discussed
only informally by Poincaré, while “Hadamard marshaled
additional evidence for the theory of unconscious processing
and tried to demonstrate the wide role of such processing in
other areas of cognition” (Weisberg 2006, 398–399). A typi-
cal example used to support this idea is Kekulé’s discovery
of benzene structure (Kekulé’s dream). While somnolent in
front of his fireplace, August Kekulé had a vision of atoms

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Table 1 Kinds of mental processes underlying the generation of hypotheses

	Parallel		Serial	
	Conscious	Unconscious	Conscious	Unconscious
Same	x	Parallelism (Poincaré)	Poincaré	x
Different	x	Associationism (Freud)	x	x

dancing one after the other and then, later, of a snake biting its tail (forming a ring). It was this dream that revealed the real structure of the benzene ring to the German chemist, according to his own report.

To provide more detail, this second viewpoint, in turn, breaks down into two branches. The separation is about the nature of mental processes, the conscious and unconscious ones, which are involved in the generation of hypotheses during the search for the solution to a problem.

A first branch maintains that conscious and unconscious processes are of the same kind: they are ‘homogeneous’ and they only differ in the way data-processing is performed. The conscious mind works in a serial way, while the unconscious works in a parallel way. Poincaré supported this idea, which is labeled *parallelism*.

By contrast, a second viewpoint states that the two mental processes are heterogeneous, they are different in kind and cannot be reduced one to the other. This hypothesis is labeled *associationism* and is supported by Freud and his followers.

In more detail, unconscious thought and its relation to conscious thought is approached along two dimensions:

- (1) the nature of the links that connect one unconscious thought to the next one. A link can employ the same processes as the conscious one or not. In the former case we have a homogeneity between conscious and unconscious thoughts, in the latter a heterogeneity.
- (2) whether the stream of thoughts are *parallel*, in the sense that you can process multiple ideas at the same time, or *serial*.

Combining these two dimensions, which are not mutually exclusive, we obtain several possible approaches to the mental processes underlying the generation of hypotheses (see Table 1).

Two of them, *parallelism* and *associationism*, as stated above, are particularly interesting for the psychological approach to scientific discovery.

The first conception states that unconscious processes are parallel and that they involve the same links as the conscious ones. The only difference is quantitative, i.e. speed, as an

outcome of parallelism: unconscious thought is much faster. Since the links that enable us to connect an idea to another are the same for conscious and unconscious processes, it follows that once the hypothesis has been generated, the person who produced it is in a position, in principle, to understand how it came out: “the unconscious does not do anything that conscious processing could not do if there were but time available” (Weisberg 2006, 395).

The second branch maintains that the links created by unconscious thought are different from those of conscious thought. That is, unconscious processing can generate associative links for reasons of which one is not aware and perhaps could not be. The difference is qualitative. So “a creative leap can come about because (1) the processing has occurred on an unconscious level, which results in the thinker’s being surprised by the sudden leap; and (2) the leap is based on connections that the person could never think of using conscious thought, which is a second source of surprise” (*Ibid.*, 389).¹

The psychology of scientific discovery has been refined several times (see for example Csikszentmihalyi and Sawyer 1995) and also popularized in movies and novels. An examination of all the different versions of it is beyond the goals of this paper. Its ‘philosophical matrix’ is clear enough. What is more interesting to discuss here are the several, critical weaknesses that the whole idea of a psychology of discovery faces. I will examine in particular two main problems:

1. the sources employed to extract the mental processes;
2. multiple discovery.

1. The source of traditional psychology of discovery and its findings is introspection, which is a conscious reconstruction of an unconscious process (the generation of a hypothesis to solve a problem). More specifically the pieces of evidence used in this case are auto-reports and interviews—such as the ones provided by Poincaré and Kekulé. But auto-reports and interviews are sources hard to trust. They are provided by individuals who “have developed their own theories of creative thinking that rely on the opportunity for unconscious processing” (Weisberg 2006, 429). Furthermore “the interviews provide evidence of the role of unconscious processing in producing illuminations”, but “the empirical support for the idea of unconscious processing in creative thinking consists of a number of anecdotal reports” (*Ibid.*).

Auto-reports and interviews not only are fragmentary and anecdotal, but they also imply “that this particular

¹ This theory, which of course stems from Freud, is a version of the out-of-mind view, as the person has no conscious awareness of, or control over, the links among ideas.

149 aspect of discovery is not available to the awareness of
150 the discoverer, and hence that the discoverer cannot be
151 a source of a reliable description of the process that pro-
152 duced the sudden image. It does not in any way imply
153 that the process is fundamentally different from other
154 processes of discovery—only that we must seek for
155 other sources of evidence about its nature” (Simon et
156 al. 1987, 328–329).

157 Bottom line: auto-reports and interviews “are not
158 adequate grounds on which to build a scientific theory
159 of creative thinking” (Weisberg 2006, 429) and they are
160 “the unwarranted source of a great deal of mysticism
161 about the discovery process. In particular, the uncon-
162 scious nature of the process is a red herring” (Simon et
163 al. 1987, 329).

164 2. A psychology of discovery does not adequately account
165 for multiple discovery (or simultaneous discovery, see
166 e.g. Lamb and Easton 1984) and neglects the role of
167 the cumulative, collective construction of hypotheses to
168 solve problems.

169 A multiple discovery happens when scientists working
170 independently of each other achieve the same discov-
171 ery at about the same time. History of science contains
172 plenty of this kind of discoveries: the formulation of cal-
173 culus by Isaac Newton and Gottfried Wilhelm Leibniz,
174 and the discovery of oxygen by Carl Wilhelm Scheele
175 and Joseph Priestley, Antoine Lavoisier (and others) are
176 probably the most recognized examples.

177 Multiple discoveries are a problem for a psychologis-
178 tic approach: if a discovery is the outcome of personal,
179 intra-psychic, and unconscious factors, it becomes hard
180 to explain how different researchers working indepen-
181 dently can discover the same thing more or less at the
182 same time. As a matter of fact an explanation of this
183 specific, but non-rare, kind of discovery requires an
184 analysis of the role of concepts such as ‘zeitgeist’, shared
185 knowledge, theories and data, which are not reducible
186 to an individual psychology of discovery: “if there is a
187 common world picture shared by the members of the
188 scientific community, certain discoveries are inevitable
189 since these discoveries are ‘in the air.’ Consequently,
190 several scientists who share the common repertoire of
191 ideas and try to solve the same problem may arrive at
192 similar solutions” (Kantorovich 1993, 186). Since the
193 list of multiple discoveries is too long to be dismissed,
194 this approach cannot provide an adequate account of sci-
195 entific discovery.

196 Moreover, “the discovery process typically is struc-
197 tured in time rather than being a momentary psychologi-
198 cal experience of the solution popping into someone’s
199 head” (Nickles 1981, 87) and, in addition “personally,

a scientist may make very few scientific discoveries, if
any, during his lifetime; most scientific discoveries are
products of collective efforts. [...] If the process extends
over a long period of time, only the final step in the
process is regarded as a discovery. Yet the contributions
of the other participants are sometimes no less impor-
tant than the contribution which constituted the break-
through” (Kantorovich 1993, 12).

Bottom line: the psychology of discovery argues that
not only is there no logic of discovery, but also that no
method is possible for it, no inferential way, and the
entire approach relies on the static, a-temporal and
obscure concept of intuition.

But many problems of a psychology of discovery
can be answered by simply noting that reasoning, argu-
mentation and inferences play a decisive role in the
process of discovery. These are rational means that
can be applied to several problems and, as such, will
often produce the same, or very similar, conclusions.
No surprise then, for instance, for multiple discovery:
“in normal science scientists who start from the same
transparent presuppositions and employ the same theory
and observational data would arrive at the same result”
(*Ibid.*, 186–187).

These and other open problems in the philosophical
matrix of discovery made necessary a deep rethinking
and reloading of it.

2 Reloading Scientific Discovery

The matrix incepted by logical empiricism and mathemat-
ical logic has been reset and then reloaded in several ways.
New approaches have been developed by questioning some
of its tenets (like the separation between context of dis-
covery and context of justification see in particular Shel-
ley 2003) and by reconsidering not only the possibility and
meaning of a logic of discovery (Hanson 1958, Simon 1987;
Nickles 1985; Cellucci 2013), but also the role played by
logic, method, inferences and even evolution in scientific
discovery (see in particular Nickles 1980a, b; Meheus and
Nickles 2009; Clement 2008; Cellucci 2013, 2017; Gro-
sholz 2007; Grosholz and Breger 2000, Ippoliti and Cellucci
2016; Nersessian 2008). I will examine three of them—the
deductive (§ 2.1), the cognitive (§ 2.2) and the evolutionary
approach (§ 2.3)²—and then I will show how the rethinking
triggered by these approaches, and their lessons, made

² This examination is not intended to be exhaustive: it analyses a few
representative works of approaches that eased a new way of account-
ing for scientific discovery.

244 possible a new theory and practice of scientific discovery—
245 a ‘scientific discovery revolution’ (§ 3).

246 2.1 Deductive Reloading

247 One way of reloading the problem of scientific discovery is
248 the deductive one. It simply maintains that there is some-
249 thing like a logic of discovery, and this is just deductive
250 logic. Of course this idea faces a straightforward objection.
251 Since deductive inference and reasoning are non-ampliative,
252 meaning that the content of their conclusion is contained
253 in the premises,³ nothing new can be inferred from them.
254 Thus, nothing new can be obtained from their application:
255 the function of deductions is to make explicit in the conclu-
256 sion pieces of information that are only implicitly contained
257 in the premises. A deduction allows us rewrite the informa-
258 tion embedded in the axioms in a way that is much more
259 understandable and testable but, from a logical viewpoint,
260 it cannot extend them: at most it is a way to establish *new*
261 logical relations between known findings, but it cannot pro-
262 duce *new* findings.

263 A response to this objection shapes the deductive view
264 of discovery (see in particular Musgrave 1988, 1989; Zahar
265 1983, 1989).

266 First, a well-known line of argument of the deductive
267 approach teases apart logical and psychological novelty
268 (see also Ippoliti 2014 on this point). It argues that even if
269 a deduction cannot be ampliative from a logical viewpoint,
270 it is ampliative from a psychological viewpoint: by deduc-
271 ing consequences we gain genuine new knowledge since the
272 truth of our premises, or postulates, is not enough to foresee
273 the truth of their consequences. We need an effort to obtain a
274 deductive conclusion from given premises or postulates—to
275 choose and combine the premises in the appropriate way—
276 and thus a deductive consequence is *new* knowledge. The
277 deductive approach supports this claim with several argu-
278 ments, such as the semi-decidability of the theories, the
279 surprise of unexpected consequences, the need of new indi-
280 viduals in deduction, and the epistemic aspect of conclusions
281 (see e.g. Dummett 1991; Hintikka 1973; Rota 1997).

282 A second line of argument of the deductive approach
283 argues that deductive reasoning provides a logic of discov-
284 ery since it is possible to rebuild ampliative inferences, such
285 as induction and analogy, in terms of deductive arguments.
286 In more detail, ampliative inferences can be reconstructed
287 as deductive inferences with suppressed premises (a kind of
288 enthymeme).

An inductive inference, argues Musgrave (1989), can be
rebuilt as a deduction with suppressed premises in the fol-
lowing way. Let us take for example the following inductive
argument I_I :

A particular raven is black
All ravens are black

In stating such an inductive hypothesis we are not walking
in dark. As a matter of fact, we move from a certain assump-
tion, such as “all ravens have some common color” (p_1). Of
course, such an assumption is domain-specific and expresses
common beliefs about our experience in a specific niche. In
turn, the assumption p_1 can be obtained from another, more
general premise p^* such as “ravens belong to a family of
kinds of birds whose members have a common color”. When
another premise, p_2 “a particular raven is black”, is added
(of course p_2 is drawn from observation), then we have the
following deductive argument D_I :

All ravens have some common color
A particular raven is black
All ravens are black

So, I_I can be converted into D_I by introducing some
appropriate assumptions or premises like p_1 or p^* . The con-
clusion of an inductive argument I_I at this point does not
constitute a novelty with respect to the premises p_1 and p_2
and follows necessarily from the them.

The same holds for analogy. An analogical inference,
argues Musgrave, can be rebuilt as a deduction with sup-
pressed premises in the following way. Let us take for exam-
ple the following analogical argument A_I , the Watson argu-
ment for the structure of DNA:

DNA has a chemical composition analogous to TMV
(tobacco mosaic virus)
TMV has a form of nucleic acid as a major chemical
constituent

The TMV molecules are helical in structure
The DNA molecule is helical in structure

A_I can be converted into a deductive argument D_2 in the
following way:

The same (similar) effects have the same (similar) causes
The chemical composition of the DNA and TMV are
similar
The structure of a TMV follows from its chemical
composition

The TMV molecule is helical in structure
The DNA molecule is structurally similar to TMV (helical in structure)

In this way we have a method of discovery based on
deductive logic (a ‘deductive heuristics’): we discover some-
thing by starting with a hypothesis that is plausible and then

³ A deductive inference is obtained by using primitive rules whereby
the content of their conclusion is literally included in their premises.
A stock example is *modus ponens* ($A, A \rightarrow B \therefore B$)— B is literally
part of the second premise. Thus a deductive reasoning, as a chain of
these basic rules, cannot expand *logically* the premises.

332 deducing consequences from it. In this sense, successful sci- 383
 333 entists are ones that employ the *right* assumptions, and a 384
 334 discovery is a deductive consequence of these assumptions 385
 335 (that are different from the ones employed by his colleagues 386
 336 at that time). 387

337 So, content-specific, suppressed assumptions are the key 388
 338 to scientific discovery and deductive reasoning is the neces- 389
 339 sary addition to them. In a sense, it is not a surprise, as the 390
 340 pivotal role of suppressed (or tacit) assumptions in scientific 391
 341 reasoning has been explicitly stated and examined (see e.g. 392
 342 Polanyi 1966). 393

343 On one hand, the presence of domain-specific suppressed 394
 344 premises in scientific reasoning would explain not only why 395
 345 it is often difficult for non-experts to understand what is 396
 346 going on inside a field of expertise, but also why the experts, 397
 347 have difficulties in explaining to non-expert their reasoning. 398
 348 These premises are invisible, or transparent, to experts as 399
 349 they employ them implicitly. They are not aware of them, 400
 350 they work as if such assumptions were self-evident, and con- 401
 351 tinually use them to fill the gaps in their inferential activity. 402

352 On the other hand, the fact that scientific reasoning is full 403
 353 of domain-specific, tacit premises allows us to explain part 404
 354 of the process of generation and selection of hypotheses in 405
 355 problem-solving. These assumptions are used to fill some 406
 356 gaps in their inferential activity and are means to *integrate* 407
 357 the data and knowledge involved in the problem and in this 408
 358 sense they shape the construction and selection of hypoth- 409
 359 eses—for example they narrow the range of possible hypoth- 410
 360 eses to be pursued. As a consequence, tacit premises offer a 411
 361 possible solution to the problem of under-determination. In 412
 362 effect, in trying to solve a problem, scientists do not have to 413
 363 choose between infinite logically possible hypotheses since 414
 364 this range is narrowed to only a few hypotheses by these 415
 365 suppressed, invisible assumptions. 416

366 Thus, suppressed assumptions play a pivotal role also in 417
 367 the way the deductive approach reloads the issue of scientific 418
 368 discovery in relation to methods, inference, and logic. In 419
 369 effect, there is no need for creativity, intuition, or genius to 420
 370 explain scientific discovery. To sum up: 421

371 logical empiricist orthodoxy is wrong: there is a logic 422
 372 as well as a psychology of invention, and there is a 423
 373 psychology as well as a logic of appraisal. Moreover, 424
 374 the logic of invention is best regarded as deductive 425
 375 logic. Finally, logical empiricist orthodoxy is also 426
 376 wrong to say that the context of invention is irrelevant 427
 377 to the context of appraisal. Consideration of inventive 428
 378 arguments yields minimal plausibility appraisals, and 429
 379 it also can tell us which facts are novel for the theory 430
 380 being considered (Musgrave 1989, 32). 431

381 Even if the deductive approach shows that at least in part 432
 382 the process of discovery can be accounted for in rational and

logical terms, the idea of “deductive heuristics” as a logic of 383
 discovery is unsatisfactory. 384

385 As concerns the first line of argument, the distinction 386
 387 between logical and psychological novelty, the arguments 388
 389 put forward to support it do not work: not only is there no 390
 391 way to logically extend our knowledge by means of deduc- 392
 393 tions from axioms but also in principle is there no need of 394
 395 any work, or effort, to get deductive consequences from a 396
 397 set of axioms, as this could be done in a mechanical way 398
 399 by using a brute force-exhaustive procedure, for instance 400
 401 the British Museum Algorithm (Newell et al. 1958). Such 402
 403 an algorithm would generate all the theorems from a set 404
 405 of premises, starting from the shorter up to the more com- 406
 407 plex ones. Even if computationally ineffective, it would not 408
 409 require any intellectual work. 410

411 As concerns the second line of argument, i.e. role of 412
 413 deductive heuristics, it does not account for a crucial step 414
 415 in discovery, that is generation of hypotheses (see Ippoliti 416
 417 2017a). It does not tell us how to get premises to use to 418
 419 derive interesting results. In addition, a deductive heuris- 419
 420 tics can be applied only when concepts or properties are 420
 421 already known and available: it is not a way for generat- 421
 422 ing new ones. Moreover, a deductive heuristics is a recon- 422
 423 struction in a deductive fashion of something that has been 423
 424 obtained in a different way. Thus, in order to be effective 424
 425 it needs a vantage point, i.e. the knowledge of the final result: 425
 426 “this is a typical example of how a process of discovery can 426
 427 be reconstructed without giving us a clue as to the method 427
 428 which might have led the discoverer to his discovery prior 428
 429 to his actually making the discovery” (Kantorovich 1994, 429
 430 8). In addition the fact that a pieces of reasoning can be re- 430
 431 built in a deductive way, does not remove the uncertainty in 431
 432 that reasoning, which is simply moved to its premises. And 432
 deductive reasoning does not and cannot justify its premises. 432

433 Of course the fact that deductive reasoning cannot 433
 434 account for the process of discovery does not imply that it 434
 435 is useless, but simply that the generation of a hypothesis—a 435
 436 crucial step in discovery—cannot be reduced to a deduction 436
 437 (on this point see also Cellucci 2013; Ippoliti 2017b). 437

2.2 Cognitive Reloading 422

423 An alternative way of reloading the problem of scientific 423
 424 discovery is the cognitive approach, and in particular a main 424
 425 branch of it: the computational approach. 425

426 This approach, incepted by Simon and Newell, draws 426
 427 upon several features of Poincaré’s account of scientific 427
 428 discovery, as it aims at providing an “examination of the 428
 429 phenomena of incubation and illumination and their expla- 429
 430 nation” (Simon 1977, 292). This theory of discovery argues 430
 431 for the so-called *nothing special view*, “which proposes 431
 432 that creative products of all sorts are brought about by our 432

433 ordinary cognitive processes, such as those involved in our
434 day-to-day problem-solving activities” (Weisberg 2006, xii).

435 Thus the cognitive approach sees scientific discovery as a
436 *problem-solving* activity and it “does not assign any special
437 role to the unconscious—or, for that matter, to the conscious.
438 It assumes, implicitly, that the information processes that
439 occur without consciousness of them are of the same kinds
440 as the processes of which the thinker is aware” (Simon 1977,
441 292). This tenet is essential to develop the most advanced
442 version of this approach, a mechanization of scientific
443 discovery—*conceptual* discovery as well as discovery of
444 *regularities*. In effect the cognitive approach argues that the
445 processes underlying a scientific discovery are ordinary (not
446 special) and can be examined, reconstructed and reproduced
447 in a rational way. In addition it argues that “human beings
448 should be looked upon as information- processing systems,
449 analogous to computers, and that the concepts underlying
450 our understanding of the functioning of computers could
451 be applied to understanding human cognition” (Weisberg
452 2006, 119–120).

453 The practical outcome of it is the BACON software pack-
454 age (Simon et al. 1987), a kind of ‘mechanical discoverer’.
455 Simon provides us with several examples of historical sci-
456 entific discovery that can be obtained by BACON, such as
457 Kepler’s third law. Fed with data on the periods of revolution
458 (P) and the distances (D) of the planets from the sun, the
459 program runs the following recursive heuristic rules:

- 460 i.) if two variables co-vary, introduce their ratio as a new
461 variable;
462 ii.) if they vary inversely, introduce their product as a new
463 variable and check if it is constant.

464 The iteration of these two simple rules is enough to obtain
465 Kepler’s third law in a purely mechanical way (see Simon
466 et al. 1987, 68–81). In a similar way, Simon shows us how
467 it is possible to account in purely computational terms for
468 many scientific discoveries, such as Boyle’s law, Ohm’s law,
469 the Law of Uniform Acceleration, Coulomb’s law, and the
470 ideal-gas law. Moreover, Simon argues that BACON would
471 be able to generate also new properties (concepts), for
472 instance ‘gravitational mass’ (Simon et al. 1987, 155–156).

473 In this way the cognitive approach provides us with an
474 answer to scientific discovery based on a rational account of
475 it: it acknowledges that the solution provided by the deduc-
476 tive approach is untenable, while providing us with a better
477 method and a theory of problems, and it shows us how infer-
478 ences and simple heuristics shape the process of generation
479 of hypotheses.

480 But while acknowledging the merits of this approach, one
481 cannot ignore all its weaknesses and the fact that this idea of
482 a mechanical discoverer appears untenable (see in particular

Nickles 1980a; Kantorovich 1993, Gillies 1996; Weisberg 483
2006; Ippoliti 2017a). 484

485 In effect BACON is capable of doing, at most, only a spe-
486 cific part of the discovery, and in a sense the (most) trivial
487 one. 488

489 First of all, BACON cannot perform a crucial step in the
490 generation of a hypothesis: the identification of relevant
491 variable and data. They are chosen and provided by the pro-
492 grammers: the software simply runs its heuristics over them.
493 This means that the *real* problem that BACON solves is
494 simply the estimation of two parameters from a set of data.
495 It provides us with an alternative solution to this elementary
496 problem and it does not discover a new generalization. 497

498 Second, BACON suffers the same weakness of the deduc-
499 tive approach: it is a reconstruction of a historical process.
500 As such, it benefits from the knowledge of a ‘backward
501 look’—the advantage of knowing that something is a prob-
502 lem, that its data can be approached by certain heuristics,
503 and that the problem is solvable. Thus, at most, we can say
504 that this software discovers in a *new solution* to a solved
505 problem. 506

507 Third and even more important, it lacks the fundamental
508 capability of producing a *really* new concept or property, or
509 better to build a conceptualization that is required to solve
510 and pose problems. Just like any software, it helps and
511 boosts human beings in computational tasks such as finding
512 regularities and making calculations.⁴ 513

2.3 Evolutionary Reloading 510

511 An alternative way of reloading the problem of scientific dis-
512 covery is the evolutionary one. This approach draws on the
513 evolutionary epistemology and on an explicit analogy with
514 biology, namely “the metaphor of an organism as a prob-
515 lem-solver” (Kantorovich 1993, 224). Thus, this approach
516 is a way of naturalizing the process of discovery, that is, to
517 explain it without assuming supernatural (in the sense of
518 not being part of natural world) entities or events (such as
519 genius). 520

521 Tellingly, while this view connects scientific discovery
522 to evolution, it denies not only the possibility of a logic, or
523 also a method, for discovery, but even the possibility of a
524 rational approach to its core—the generation of hypotheses.
525

526 In particular this view makes use of the mechanism
527 of evolutionary process, namely the processes of blind
528

⁴ Simon (Kulkarni and Simon 1988) relaxed the structures of
BACON in later programs such as KEKADA, which, unlike BACON
that “was concerned mainly with the ways in which theories could
be generated from empirical data, with little or no help from theory”
(Kulkarni and Simon 1988, 140), tries to deal with issues such as the
question of where the data came from, the processes of designing
experiments and programs of observation.

526 variations and selective retention (BV + SR), as an explanation
527 for discovery. Such a reading of the process of discovery
528 employs a dynamic view, according to which growth of a
529 theory can be seen as the growth of an organism: in effect
530 “both a research program, and an organism are engaged in
531 problem-solving. A research program solves, for instance,
532 problems arising from the need for adjusting its basic ideas
533 [...] to new observational data. Similarly, an organism adapts
534 to new environmental conditions on the basis of its genetic
535 makeup” (*Ibid.*).

536 In effect the BV + SR mechanisms seem to be “incredibly
537 creative problem solvers” (Nickles 2009, 193) and “biological
538 evolution can be interpreted, metaphorically, as an
539 innovative problem-solving process” (*Ibid.*, 182). One of
540 advantages here is the fact that they seem to offer a way to
541 get *more* knowledge from *less* knowledge: the recombination
542 and blind variation of few simple items seems capable of
543 producing something (more) complex and articulated.

544 Two of the main consequences of an evolutionary
545 approach to scientific discovery, or better of the idea that
546 discovery is a blind variation, are that it cannot be method-
547 governed and that science has no predetermined goal. These
548 two features, of course, are worrying since they seem to sup-
549 port the thesis that discovery not only is unintentional, but
550 also that it happens by ‘chance’. An answer to this objection
551 (see e.g. Kantorovich and Ne’eman 1989) is “that the coun-
552 terpart of blind biological mutation in science be interpreted
553 as serendipitous discovery, which means that scientists
554 proceed in a methodical or guided way, even though their
555 final discovery may solve a problem they had not originally
556 intended to solve” (Kantorovich 1994, 19). An outcome of
557 this way of modeling scientific discovery is that the human
558 mind is the ‘host’ of the process of blind variation and selec-
559 tive retention and “these processes can be categorized as
560 unintentional or involuntary. The incubation process is well
561 known, but is not well understood. It can be explained by
562 the model of natural selection which is applied to mental
563 elements created quasi-randomly in the discoverer’s mind;
564 the discoverer hosts the process in his mind, so to speak”
565 (*Ibid.*, 20). This idea of our mind as a mere container of
566 a combinatorial-selective process enables a surprising ver-
567 sion of the evolutionary approach, that is a computational
568 one, which uses genetic algorithms (see Koza 1992, 1994;
569 Koza et al. 1999). In this particular sense, the evolutionary
570 approach does defend an idea of a method for discovery,
571 even if a limited one. Here rationality is at the beginning
572 and at the end of a process, which takes place by combining
573 items in random way: humans can set-up the process, wait
574 for the outcomes and select one of them. So in this sense,
575 against Popper and Campbell, an evolutionary approach

would not be the antithesis of method and to some degree it
576 could be methodized.⁵ 577

578 Nonetheless, the endpoint of this evolutionary approach
579 is hard to defend. 580

581 First, the idea that serendipity⁶ can be used as a general
582 model to account for scientific discovery seems untenable
583 for at least two reasons—one quantitative and one quali-
584 tative. On one hand, the discoveries made in that way are
585 few, and, at most, they are an epiphenomenon. On the other
586 hand, the very notion of ‘discovery by chance’ does not
587 explain precisely the cases that it employs to support its
588 own view. Fleming, Columbus, or Poincaré, did not discover
589 ‘by chance’, but the discovery was made after they used a
590 lot inferences, strategies, and plans—all rational tools. Of
591 course the final product was in a sense unexpected, but it was
592 found after many options had been explored and rejected by
593 the discoverer or others (e.g. in the case of Poincaré’s math-
594 ematical discovery), or after several controlled trials (e.g. in
595 the case of Fleming’s discovery of penicillin), or simply the
596 final discovery was not excluded from the very beginning
597 (e.g. Columbus in trying to circumnavigate the earth could
598 not exclude that other lands could be encountered).

599 Second, evolution can be used to support a different and
600 more rational account for scientific discovery. In effect, it
601 can be argued that evolution shows us that biological pro-
602 cesses are not simply a kind problem-solving, but *inferential*
603 problem-solving also at a very basic and primitive level (see
604 von Helmholtz 1866; Gregory 1970, 1980). Vision would
605 provide a stock example in this sense. Following this tra-
606 dition and its findings, our visual apparatus can be recon-
607 structed in inferential term, as it continually would make
608 hypotheses to solve problems: it is not a camera, but an
609 inferential machine (see e.g. Cellucci 2013, 2017a). These
610 inferences are ampliative, unconscious, non-propositional
611 and built-in by evolutionary adaptation: they processes data
612 (light beams) and generate hypotheses using non-deductive
613 rules on the base of past experience. So, here there is not a
614 neat distinction between generation and justification of a
615 hypothesis, as the data to generate and verify a hypothesis
616 can be the same. In addition these inferences and hypotheses
617 are domain-specific: a change of environment or niche can
618 lead them to error.

619 This way of naturalizing scientific discovery, which
620 explicitly connects a method-governed approach to discov-
621 ery and evolution, changes again our theories of discovery
622 and it paves the way for the construction of a variety of
623 rational tools and *rules* to advance knowledge: it harmonizes
624 theory and practice of scientific discovery in a *new* way and

⁵ I would like to thank Tom Nickles for a clarification of this point. 5FL01

⁶ See also (Gillies 2014 and Roberts 1989). 6FL01

624 opens the door to a *logic* of discovery. It initiates a scientific
625 discovery revolution.

626 3 Scientific Discovery Revolutions

627 This new view on theory and practice of scientific discovery
628 has its roots in Lakatos' work on heuristics (Lakatos 1976)
629 but has expanded, revised and improved it a lot.

630 It borrows from Lakatos' work and from the several
631 reloaded versions of the issue the idea that scientific discov-
632 ery is problem-solving,⁷ and argues that it can be accounted
633 for in an inferential way. Moreover it reconciles evolution
634 and logic, since it argues that the way we discover something
635 is by using a kind of reasoning that has been built-in our
636 cognitive and biological make-up by adaptation (see Cel-
637 lucci 2013). This new way of characterizing scientific discov-
638 ery not only removes the separation between the genera-
639 tion and the testing of a hypotheses (as heuristic procedures
640 play a role also in testing and selecting a hypothesis), but it
641 also provides a unifying approach to problem-solving and
642 problem-finding.

643 I will examine its theory of discovery and then I will
644 look at the practices suggested by its way of conceptualizing
645 scientific discovery.

646 As concern the former, it maintains that discovery is a
647 *natural* process that has a *method*, which is bottom-up, and
648 a *logic* with *rules*. By methods it means simply a means
649 to an end, a way of pursuing a certain aim. It is provided
650 by a specific version of the analytic method, which offers
651 a very general bottom-up framework for solving problems
652 (see Cellucci 2013). More precisely, we start looking for a
653 hypothesis that is a sufficient condition for solving the prob-
654 lem—that implies a solution to the problem. This hypothesis
655 is obtained from the bottom, i.e. from the problem and pos-
656 sibly other data already available, by the application of some
657 non-deductive rules. The only requirement is that it has to
658 be plausible.⁸ Then the hypothesis undergoes a plausibility
659 test and, in positive case, is provisionally accepted. In turn
660 this hypothesis is a new problem, which must be solved and
661 its solution is sought in the same way.

662 Of course in order to work, and to produce a logic of
663 discovery, this framework must be endowed with proced-
664 ures for the generation of hypotheses. These procedures
665 are non-mechanical, as “the purpose of a logic of discovery
666 is not to dispense with the need for intelligence by use of an

algorithmic method, but rather to expand natural intelligence
providing it with heuristic means—means capable of guiding
natural intelligence, albeit not infallibly” (Cellucci 2017a,
143–144). In more detail, these procedures are heuristic in
kind, for they do not guarantee a solution of the problem:
they are ampliative and as such they can lead to erroneous
conclusions. Nonetheless, they are a *genuine* way to con-
struct new hypotheses in a rational way.

Here comes the practical, and in some ways more impor-
tant, side of this new conceptualization of scientific discov-
ery. In this sense ‘scientific discovery revolution’ argues that
we have to stop thinking about a *theory* of discovery, and
to concentrate on the investigation of procedures to be *used*
to generate new hypotheses to solve problems. The study,
production and classification of heuristic procedures are the
frontier of a logic of discovery, and not surprising they have
been improved a lot recently (see in particular Cellucci 2013
ch. 20–21; Jaccard and Jacoby 2010; Ippoliti 2017b; Ippoliti
and Cellucci 2016).

In essence these heuristic procedures are tools for *model-
ling* the research space of hypotheses—what Simon defines
the ‘problem-space’. To produce a new hypothesis we have
to combine ideas and concepts in many ways and this will
generate a combinatorial explosion in the problem-space. A
heuristic procedure is a means to handle such a combinato-
rial space: it is a tool to build such a space, as this space
essentially depends on the existing knowledge and the avail-
able data. Such a space is determined not only (i) by the way
the data are processed using a heuristics, but also (ii) by the
way data and the corpus of existing knowledge change over
time. The application of a heuristic procedure can reduce as
well as expand a problem-space.

A heuristics can be classified in several ways, start-
ing from the well-known distinction between *positive* and
negative heuristics (see Lakatos 1976).⁹ A further useful
distinction is the one that tells apart *primitive* and *derived*
heuristics, or *generative* and selective *heuristics* (see Ippoliti
2017b). The primitive are analogies, disanalogies, induc-
tions and their combinations (e.g. analogies between analo-
gies). Any heuristic procedure is a variant, combination or
juxtaposition of these primitive heuristics, for it *presup-
poses* them. They enable us to build a more articulated and
complex (derived) heuristics, e.g. metaphors or scenario-
building. The set of derived heuristics is *open*, as new rules
can be produced as new problems are solved, nonetheless
we can classify them into several classes: inversion heu-
ristics, heuristics of switching, scenario building, thought

⁷ This idea goes back to Plato, and Aristotle (see Quarantotto 2017)
and has been re-proposed also recently (Laudan 1977, 1981; Nickles
1981; Cellucci 2017b).

⁸ Plausible here, following Aristotle's notion of *andoxa*, simply
means that the arguments for the hypothesis are ‘stronger’ (in quality)
than those against it on the basis of the existing knowledge.

⁹ A positive heuristics guides us in the construction of admissi-
ble paths during the search of a solution for a problem. A negative
heuristics prevents us from building certain paths—by blocking the
modus tollens on a specific part of the theory.

714 experiments, the analysis of extreme cases, the analysis of a
 715 deviant case amongst others (see Jaccard and Jacoby 2010;
 716 Ippoliti 2017b). For example the heuristics of switching¹⁰ is
 717 based on a specific change of viewpoint, which we get by
 718 switching from one order of analysis to another one. Thus
 719 it is based on the study of explicit analogies and disanalogies.
 720 ¹¹ Moreover these heuristic procedures cover both sides
 721 of problem-solving, namely to *solve* and *pose* problems.¹²

722 Thus this new, inferential, approach has revolutionized
 723 not only the way of theorizing discovery but also, and more
 724 importantly, the ways of practicing it in a rational and more
 725 systematic manner. We have now a better, finer-grained
 726 understanding of methods and toolboxes to use as a guide
 727 to discover something new. And we can improve them. This
 728 is a remarkable contribution of philosophy of science to sci-
 729 ence, as it puts the former in a position not only to interpret
 730 what scientists do, but also to provide and improve tools that
 731 they can employ in their activity.

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735 References

736 Cellucci C (2013) Rethinking logic. Springer, Dordrecht
 737 Cellucci C (2017a) Rethinking knowledge. Springer, Dordrecht
 738 Cellucci C (2017b). Is mathematics problem solving or theorem prov-
 739 ing?. *Found Sci* 22(1):183–199
 740 Clement JJ (2008) Creative model construction in scientists and
 741 students. The role of imagery, analogy, and mental simulation.
 742 Springer, New York
 743 Csikszentmihalyi M, Sawyer K (1995) Creative insight: the social
 744 dimension of a solitary moment. In: Steinberg RJ, Davidson JE
 745 (eds) *The nature of insight*. MIT Press, Cambridge, pp 329–361
 746 Dummett M (1991). *Frege. Philosophy of mathematics*. Duckworth,
 747 London
 748 Einstein A (1958) A testimonial. In: Hadamard J (ed) *The psychol-
 749 ogy of invention in the mathematical field*. Dover, Mineola,
 750 pp 142–143
 751 Frege G (1960) *The foundations of arithmetic. A logic-mathematical
 752 enquiry into the concept of number*. Harper, New York
 753 Gigerenzer G, Todd P, ABC Group (1999). *Simple heuristics that make
 754 us smart*. Oxford University Press, New York
 755 Gillies D (1996) *Artificial intelligence and scientific method*. Oxford
 756 University Press, Oxford
 757 Gillies D (2014) Serendipity and mathematical logic. In: Ippoliti E,
 758 Cozzo C (eds) *From a heuristic point of view*. Cambridge Scholars
 759 Publishing, Newcastle upon Tyne, pp 23–39

Gregory RL (1970) *The intelligent eye*. Weidenfeld & Nicolson, London 760
 Gregory RL (1980) Perceptions as hypotheses. *Phil Trans R Soc Lond B* 290:181–197 761
 Grosholz E (2007) *Representation and productive ambiguity in math- 762
 ematics and science*. Oxford University Press, New York 763
 Grosholz E, Breger H (eds) (2000) *The growth of mathematical knowl- 764
 edge*. Springer, Dordrecht 765
 Hadamard J (1958) *The psychology of invention in the mathematical 766
 field*. Dover, Mineola 767
 Hanson N (1958) *Patterns of discovery: an inquiry into the conceptual 768
 foundations of science*. Cambridge University Press, Cambridge 769
 Hintikka J (1973) *Logic, language-games and information*. Oxford 770
 University Press, Oxford 771
 Ippoliti E (2014) Reasoning at the frontier of knowledge. In: Ippoliti E
 (ed) *Heuristic reasoning*. Springer, Basel, pp 1–10 772
 Ippoliti E (2017a) Building theories. The heuristic way. In: Dank D,
 Ippoliti E (eds) *Building theories*. Springer, Berlin 773
 Ippoliti E (2017b). Heuristic logic. A kernel. In: Danks D, Ippoliti E
 (eds) *Building theories*. Springer, Berlin 774
 Ippoliti E, Cellucci C (2016) *Logica*. Egea, Milan 775
 Jaccard J, Jacoby J (2010) *Theory construction and model-building*.
 Guilford Press, New York 776
 Kantorovich A (1993) *Scientific discovery: logic and tinkering*. State
 University of New York Press, Albany 777
 Kantorovich A (1994) Scientific discovery: a philosophical survey.
Philosophia 23(1–4):3–23 778
 Kantorovich A, Ne’eman Y (1989). Serendipity as a source of evolu-
 tionary progress in science. *Stud Hist Philos Sci* 20(4):505–529 779
 Koza J (1992) *Genetic programming: on the programming of comput-
 ers by means of natural selection, vol I*, MIT Press, Cambridge 780
 Koza J (1994) *Genetic programming II: automatic discovery of reus-
 able programs*. MIT Press, Cambridge 781
 Koza J, Bennett III, Andre F, Keane D, M (1999) *Genetic programming
 III: Darwinian invention and problem solving*. Morgan Kaufmann,
 San Francisco 782
 Kulkarni D, Simon H (1988). The processes of scientific discovery:
 the strategy of experimentation. *Cognitive Science*, 12:139–175 783
 Lakatos I (1976) *Proofs and refutations: the logic of mathematical
 discovery*. Cambridge University Press, Cambridge 784
 Lamb D, Easton SM (1984). *Multiple discovery: the pattern of scien-
 tific progress*. Avebury Publishing Company, London 785
 Laudan L (1977) *Progress and its problems*. University of California
 Press, Berkeley 786
 Laudan L (1981) A problem-solving approach to scientific progress.
 In: Hacking I (ed) *Scientific revolutions*. Oxford University Press,
 Oxford, pp 144–155 787
 Meheus J, Nickles T (eds) (2009) *Models of discovery and creativity*.
 Springer, Dordrecht 788
 Murray P (1989) *Genius: the history of an idea*. Blackwell, Oxford 789
 Musgrave A (1988). Is there a logic of scientific discovery?. *LSE Q*
 2(3):205–227 790
 Musgrave A (1989) Deductive heuristics. In: Gavrovlou K, Goudaroulis
 Y, Nicolacopoulos P (eds) *Imre Lakatos and theories of scientific
 change*. Kluwer, Boston, pp 15–32 791
 Nersessian N (2008) *Creating scientific concepts*. MIT Press, Cam-
 bridge (MA) 792
 Newell A, Shaw JC, Simon HA (1958) Elements of a theory of human
 problem solving. *Psychol Rev* 65(3):151–166 793
 Nickles T (ed) (1980a) *Scientific discovery: logic and rationality*.
 Springer, Boston 794
 Nickles T (ed) (1980b) *Scientific discovery: case studies*. Springer,
 Boston 795
 Nickles T (1981) What is a problem that we may solve it? *Synthese*
 47(1):85–118 796
 797
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¹⁰ Stock examples are: change of *unit* of analysis, change of *level* of analysis, focus on *processes* vs focus on *variables*.

¹¹ An *atomistic* view of heuristics, that there is an ultimate, base set from which all others can be compounded, has also been put forward by Gigerenzer and Todd (see Gigerenzer et al. 1999).

¹² There are many examples of problems generated by heuristic procedures, like Poincaré’s conjecture.

- 825 Nickles T (1985) Beyond divorce: current status of the discovery
826 debate. *Philos Sci* 52(2):177–206
- 827 Nickles T (2009) The strange story of scientific method. In: Meheus
828 J, Nickles T (eds) *Models of discovery and creativity*. Springer,
829 Dordrecht, pp 167–207
- 830 Poincaré H (1908). *L'invention mathématique*. *Enseignement mathé-*
831 *matique* 10:357–371
- 832 Polanyi M (1966) *The Tacit Dimension*. Routledge, London
- 833 Popper K (1961) *The logic of scientific discovery*. Science Editions,
834 New York
- 835 Quarantotto D (2017) Aristotle's Problemata style and aural textuality.
836 In: Polansky R, Wians W (eds) *Reading Aristotle*. Brill, Leiden,
837 pp 75–122
- 838 Roberts RM (1989) *Serendipity: accidental discoveries in science*.
839 Wiley, Hoboken
- 840 Rota GC (1997) *Indiscrete thoughts*. Birkhäuser, Boston
- 841 Shelley C (2003) *Multiple analogies in science and philosophy*. John
842 Benjamins B.V, Amsterdam
- Simon H (1977) *Models of discovery*. Dordrecht, Reidel 843
- Simon H (1987) Is scientific discovery a topic in the philosophy of
844 science? In: Rescher N (ed) *Scientific inquiry in philosophical*
845 *perspective*. University Press of America, Lanham, pp 1–15 846
- Simon H, Langley P, Bradshaw G, Zytkow J (1987) *Scientific discov-*
847 *ery: computational explorations of the creative processes*. MIT
848 Press, Boston 849
- Simonton DK (1988) *Scientific genius: a psychology of science*. Cam-
850 bridge University Press, Cambridge 851
- von Helmholtz H (1866) *Concerning the perceptions in general*. In:
852 *Treatise on physiological optics*, vol III. Dover, New York, pp
853 1–37 854
- Wallas G (1926) *The art of thought*. Cape, London 855
- Weisberg R (2006) *Creativity: understanding innovation in problem*
856 *solving, science, invention, and the arts*. Wiley, Hoboken 857
- Zahar E (1983). *Logic of discovery or psychology of invention?*. *Brit-*
858 *ish J Philos Sci* 34:243–261 859
- Zahar E (1989) *Einstein's revolution*. Open Court, La Salle 860

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