

The division of labour in science: the tradeoff between specialisation and diversity

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Economics is a typical resource for social epistemology and the division of labour is a common theme for economics. As such it should come as no surprise that the present paper turns to economics to formulate a view on the dynamics of scientific communities, with precursors such as Kitcher (1990), Goldman and Shaked (1991) and Hull (1988). But although the approach is similar to theirs, the view defended is different. Mäki (2005) points out that the lessons philosophers draw from economics can go either way depending on the model chosen. Thus, the aims of this paper are (1) to illustrate this flexibility by proposing an alternative model which assumes increasing returns to adoption in science rather than the decreasing returns present in the aforementioned contributions; and (2) to outline the implications of this view for scientific pluralism and institutional design.

Keywords: economic epistemology; division of labour; increasing returns; network industries; scientific pluralism; institutional design

1 Introduction

The subject of this paper is the division of labour in science. Should scientists diversify or specialise? The larger a consensus within a group of scholars, the more scholars can reap the benefits from dividing labour and the specialisation it allows; the more dissensus, the richer the diversity of views. Both specialisation and diversity have been attributed epistemically beneficial features. However, they cannot simply be maximised simultaneously because they are inversely related; more diversity means fewer possibilities for specialisation and vice versa. As such, a *tradeoff* obtains between the two. This gives rise to a number of related questions. Is there an optimal distribution we should aim for in science? Is it specific for a discipline, domain, group, question, . . . or science in general? Does it change throughout history and if so, what are its historical and social determinants? *Can* we change an existing balance, *should* we do it and *how* can it be done? And who are *we* anyway?

Any group, academic or other, occupies at any time a certain point in between these two extremes of specialisation and diversity. To determine whether that point is also the *right* one, an answer is needed to the question of how to distribute labour. This means that the relevance of this question is both theoretical and practical. The broader implications lie, theoretically, with scientific pluralism (i.e. the normative endorsement of a plurality of views in science). An answer to the question how to distribute labour in science can be seen as a measure for the desirability of pluralism. Also, analysing pluralism from the

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perspective of the tradeoff between specialisation and diversity provides more clarity on the *kind* of pluralism that is preferable. The practical significance is that it can yield guidelines for institutional design and arguments to be used by (science) policymakers.

The paper is organised as follows. Section 2 explains *what* this tradeoff is and how it relates to science. Section 3 introduces increasing returns and relates them to scientific knowledge. Section 4 explores the analogy between science and network industries; section 5 then offers an application of this analogy to the economics discipline. The view of science as a network industry is used in section 6 to formulate an answer to the question which position on the tradeoff *should* be preferred. Finally, section 7 examines how the desiderata from the previous section *can* be achieved and section 8 concludes.

2 Specialisation and diversity

The question of the division of labour in science has been characterised by Helen Longino as ‘the question whether and when to pursue research that calls a community consensus into question or to pursue research that extends the models and theories upon which a community agrees’ (Longino 2006). In other words, the question whether science should *specialise* or *diversify*. The benefits of diversity have been widely recognised, for example by Philip Kitcher, who states that ‘we sometimes want to maintain cognitive diversity even in instances where it would be reasonable for all to agree that one of two theories was inferior to its rival, and we may be grateful to the stubborn minority who continue to advocate problematic ideas’ (Kitcher 1990, p. 16). Similarly, Paul Feyerabend wrote ‘that progress can be brought about only by the active interaction of different “theories”’ (Feyerabend 1970, p. 203). More generally, the merits of diversity spring from the fact that alternatives provide each other with valuable criticism (Mill, Popper, Feyerabend, Longino) and enable a division of labour (Hull 1988; Kitcher 1990; Goldman 1991). In the economy, the presence of a diversity of firms, products and preferences is said to bring about competition, higher productivity and a greater collective benefit.

On the other hand, scholars like Thomas Kuhn hold that the main emphasis should lie with specialisation: ‘At least for the scientific community as a whole, work within a well-defined and deeply ingrained tradition seems more productive of tradition-shattering novelties than work in which no similarly convergent standards are involved.’ (Kuhn 1977, p. 234). And of course the economic benefits associated with specialisation have become received wisdom ever since Adam Smith’s tale of the pin factory.¹ According to Smith, specialisation is beneficial because it prompts a strict definition of tasks, leads to the development of more specific tools and allows workers to achieve a higher level of personal skill.

In sum, both specialisation and diversity have been attributed epistemically benign features. But, perhaps disconcertingly, they are inversely related to each other: more diversity means fewer possibilities for specialisation and vice versa. However, there is an obvious sense in which specialisation and diversity are not at odds at all. For example at an assembly line, a paradigm example of specialisation, there is a large amount of diversity because the workers perform different, complementary tasks. Specialisation here boosts diversity rather than impedes it. The reason why the tradeoff fails to obtain here is because two levels are being confused: the level of the standard which enables the division of labour (the assembly line) and the higher level which sees *across* such standards (cf. *infra* for more on standards). Within the standard, at the level of the assembly line itself, there is indeed diversity; but seen from the higher level, there is only one standard at work. It is exactly because everyone shares the same standard that the division of labour at the

99 assembly line is so efficient. If different standards were to operate simultaneously, there
 100 would be more diversity but productivity would suffer. So both specialisation and
 101 diversity must explicitly be situated at the inter-standard level in order to make sense of the
 102 tradeoff as it was introduced. A discipline is then specialised when the community's
 103 research efforts are concentrated on a small part of the available standards; it is diverse
 104 when research effort is equally distributed among standards.

105 Given that both are said to be epistemically beneficial and cannot be maximised
 106 simultaneously, a balance has to be struck. Somehow specialisation and diversity need to be
 107 traded off against each other. This problematic relationship has not remained unnoticed.
 108 Thomas Kuhn calls it 'the essential tension'² between traditionalism and iconoclasm; Paul
 109 Feyerabend states that the interplay between what he calls 'the principle of tenacity' and
 110 'the principle of proliferation' (roughly equivalent to what I call specialisation and
 111 diversity) is an essential feature of the actual development of science.³ Larry Laudan (1984)
 112 pointed out that no satisfactory account has been developed which simultaneously explains
 113 consensus and divergence in science.⁴ But even going back to Adam Smith leads us to a
 114 corollary of this tension,⁵ namely the tension between the already mentioned pin factory
 115 and that other famous concept of his: the invisible hand. Smith uses the example of the pin
 116 factory to argue that the amount of specialisation is limited by the extent of the market; the
 117 bigger the market, the more specialisation and the more economies of scale. The ideal
 118 factory size is thus one that dominates the entire market, i.e. a monopoly. On the other hand,
 119 the invisible hand which is necessary for optimal allocation can only function properly in a
 120 world of perfect competition. As such, the pin factory and the invisible hand are, strictly
 121 speaking, inconsistent.⁶ Actually they might be compatible under special conditions, but the
 122 increasing and decreasing returns, for which they are a metaphor, are not. Increasing returns
 123 set in motion a virtuous circle that make it tend toward monopoly. On the other hand, in a
 124 context of decreasing returns, optimal size is usually smaller than the entire market, leaving
 125 room for alternatives. Because of the relation between the assumption of increasing and
 126 decreasing returns on the one hand and the tension between specialisation and diversity on
 127 the other hand, the assumption about returns to adoption will play a key role in this paper.

128 129 130 **3 Knowledge and increasing returns**

131 Increasing returns are best known in the form of increasing returns to scale, a part of
 132 standard economic theory. As a company gets bigger it can realise a higher level of
 133 efficiency. These gains are linear and affect only the company involved. Returns to scale
 134 are a common occurrence in the literature on scientific development. For example the
 135 theory of the cycle of credibility, in Latour and Woolgar (1986), which demonstrates the
 136 mechanism of a virtuous cycle of obtaining credit, transforming it to more resources for
 137 new research and hence obtaining even more credit. These benefits are linear and restricted
 138 to the individual. By contrast, networked increasing returns are exponential and the gains
 139 are spread throughout the entire network. A telephone network is a good example.
 140 A telephone is worthless without other people to call; its value increases as more people
 141 get one. As these networks grow, their value grows exponentially and for all adopters; one
 142 extra adopter enables two calls, two extra gives four additional connections, etc. Crucially,
 143 the value of the network depends on the number of agents adopting a certain standard.

144 Note that the level of analysis is here the activity *between* different communities
 145 adopting different standards. Within a certain community there might well be free
 146 competition among scientists, but across communities this competition is thwarted by
 147 transaction costs involved in switching standards. The point is that if one takes these

standards seriously, the problem of the division of labour changes from the dynamics within a consensual community (as in Kitcher) to the dynamics between standards competing for adoption. From an institutional design perspective, this is the relevant level because it's here that science policy goals can be pursued.

In economics, the main difference between the pin factory and the invisible hand is that the former is driven by increasing returns and the latter by decreasing returns. The bigger the pin factory, the more specialisation is possible and the higher the profit. Hence there is no optimal company size; the best size is determined by the extent of the market. On the other hand, the invisible hand relies on competitive equilibrium, with an optimal company size due to decreasing returns. Although there are plenty of instances in the economy characterised by increasing returns, such as railways, telecoms, energy,⁷ a lot of contributions in economic theory have traditionally presupposed decreasing returns. This is perhaps because decreasing returns make it possible to find a single, optimal equilibrium and enable the use of clean and powerful mathematical tools. Increasing returns, on the other hand, entail multiple and possibly suboptimal equilibria, path-dependence and sticky prices, and cause economies to lock into inefficient technologies. In other words, assuming decreasing returns is 'the path of least mathematical resistance' (Krugman 1994, p. 4).

Although economics is currently undergoing an 'increasing returns revolution', philosophers of science drawing on economics have tended to follow the most accepted strands of traditional theory and their assumptions, including the decreasing returns one: models of scientific activity such as Kitcher (1990), Goldman and Shaked (1991) and Hull (1988) presuppose that scientific activity is characterised by decreasing returns to adoption. Is this based on an explicit argument for the prevalence of decreasing returns in science or have they just turned to economics in search of tools and picked those tools which were best elaborated? Whether increasing or decreasing returns are prevalent in science is ultimately an empirical matter, but in the meantime it makes sense to develop an alternative view that questions the implicit assumptions of those previous models and explores their consequences. As Mäki (2005) has pointed out, the models in economics from which social epistemology can draw can be used to make different and incompatible points about science. This means Kitcher's lesson can be turned upside down. To this effect, I will point out how social aspects of science can cause specialisation rather than diversification depending on a different assumption about returns to adoption. In particular, in Kitcher (1990), the number of adopters negatively affects the number of future adopters because it limits agents' chances to be the first to come up with the solution. But if science is seen as a network activity (distributed), the stronger – i.e. the larger – the network, the more agents will want to be compatible with it. Hence the number of adopters *positively* affects the number of future adopters.

So what would such an alternative model look like? The value of the (product of the) network does not primarily depend on scarcity, what it is made of or who made it, but on how many people use it. Consequently, seeing scientific communities as networks means they increase exponentially in strength every time a node is added to the network. As such, the addition of one scientist to a scientific community wouldn't decrease the other scientists' prospects of success (as in Kitcher) but would increase the prospect of success for the community exponentially. It would be a model of scientific activity under increasing returns to adoption.

In a series of papers, Brian Arthur has tried to find a formal way of capturing the dynamics of systems exhibiting such increasing returns to adoption. Arthur (1989) points out that many increasing-return problems fit a general nonlinear probability schema. In Arthur (1994), the general idea is laid out as follows:

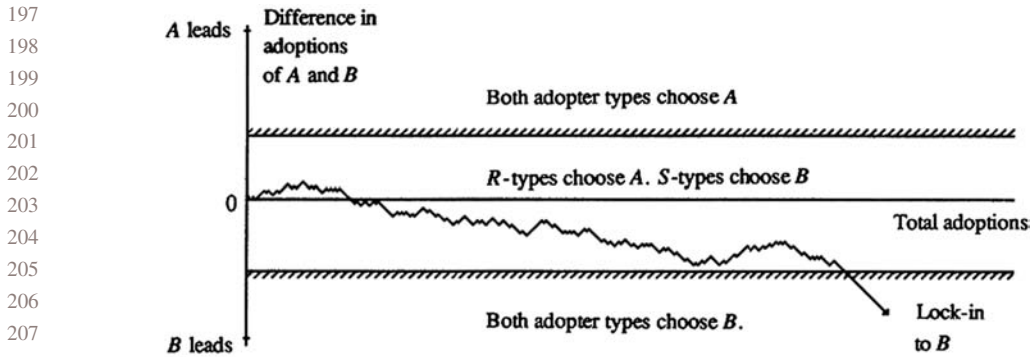


Figure 1. Increasing returns adoption: a random walk with absorbing barriers.

Source: Adopted from Arthur (1989), p. 120.

Q2 It can be pictured by imagining a table to which balls are added one at a time; they can be of several possible colors – white, red, green or blue. The colour of the ball to be added next is unknown, but the probability of a given color depends on the current proportions of colors on the table. If an increasing proportion of balls of a given color increases the probability of adding another ball of the same color, the system can demonstrate positive feedback. (Arthur 1994, p. 6)

The result of such a system is a random walk with absorbing barriers: the ratio of colours varies randomly but if the ratio of any colour reaches a certain level (the barrier) it will never get under that level afterwards. In other words, the outcome is locked in.

A paradigm for such models has been set in Arthur (1989), in which a simple example is provided in which two technologies, A and B, compete for adoption. There are also two types of agents: R-types choose A and S-types choose B. Agents are thrown into the game with a 0.5 probability of being an R-type or an S-type. This probability changes once the amount of adopters of A or B reaches a certain threshold, after which all new agents adopt to the dominant technology irrespective of their being an R- or S-type. The resulting dynamic is one of a random walk with absorbing barriers. Interestingly, it is certain that agents will eventually lock into one of the two technologies, but which one is impossible to predict; it is different every time the game is played. More generally, Brian Arthur notes that the characteristics of the dynamics of this game will differ in crucial respects from the dynamics of games assuming decreasing returns: it will exhibit multiple equilibria, non-predictability, potential inefficiency, inflexibility and non-ergodicity.

4 Science as a network industry

The main focus of the present paper lies in the tradeoff between specialisation and diversity and its implications for pluralism. For a detailed formal model applying increasing returns to scientific activity, see De Langhe and Greiff (in press). In this paper the emphasis is on the development of the analogy between scientific activity and network industries.

Hans-Werner Gottinger gives the following definition of a network industry:

Network industries can be defined as those where the firm or its product consists of many interconnected nodes, where a node is a unit of the firm or its product, and where the connections among the nodes define the character of commerce in the industry. (Gottinger 2003, p. 1)

246 In a similar sense, I take a scientific community to consist of a network of nodes. The
 247 nodes are individual contributions (a paper, a book, introduction of a new idea). A
 248 necessary condition for these nodes to form a network (a scientific community) is that
 249 there is a *standard* which makes the nodes compatible with each other. Each time a
 250 scientist makes a contribution, the contribution falls under a certain standard and the
 251 scientist can be said to ‘adopt’ to that standard at that point. Adopters to a certain standard
 252 form a community and different communities based on different standards then compete at
 253 the level of the discipline. It is important to note that the very possibility of a division of
 254 labour in science crucially depends on the presence of such a standard. Without a shared
 255 standard within which different actors can perform different tasks but still work on a
 256 common project it would be impossible for scientists to coordinate their research efforts.
 257 Standards are a condition of possibility for the aggregation of individual contributions and
 258 for accumulation of the results over time. As such the topic of division of labour cannot be
 259 addressed without at least implicitly presupposing the presence of a standard. This paper
 260 aims to make this presupposition explicit and explore its consequences by making the
 261 analogy between the dynamics of scientific activity within a discipline and the dynamics
 262 of standards competing for adoption.
 263

264 There are a number of good *prima facie* reasons for exploring this analogy. First, as
 265 emphasised in recent contributions such as Giere and Moffatt (2003) and Hutchins (1995),
 266 scientific activity is a distributed effort, much like the individual nodes forming the
 267 network featured in the definition of a network industry. Secondly, scientists seem to
 268 cluster their ideas around certain basic concepts and key theoretical assumptions, for
 269 example those called ‘paradigms’ by Kuhn (1962); these act as *standards* with which
 270 scientists comply. Thirdly, information is characterised by increasing returns. This has
 271 important effects on companies that produce them, for those companies function
 272 differently from other, mainly industrial, companies. Fourthly, a bigger network also
 273 means more feedback, more conferences to go to, more journals to publish in, more jobs to
 274 apply for, etc. Networks do seem to matter in the academic world. A fifth factor is that the
 275 incentive structures at universities tend to reward research that has been successful in the
 276 past. Those rewards often come in the form of more research means, enabling the
 277 successful researchers to be even more successful in the future. This has even grown with
 278 the increasing importance of bibliometric methods, which has made joining a larger
 279 network even more compelling. A simple example can illustrate this powerful feature of
 280 networks. A discipline has two research communities (RC). RC1 has three members, RC2
 281 has four. Now let’s say all members publish a paper and cite all other members in that
 282 paper. In RC1 this will give each member six citations, in RC2 the extra member *doubles*
 283 the number of citations (12). This provides a strong incentive for a newcomer to join RC2,
 284 which would in effect make the discrepancy between RC1 and RC2 even stronger (20 vs
 285 6). As such, the larger the network with which a scientist’s contribution is compatible, the
 286 larger its impact.
 287

288 Using Arthur’s paradigm as a starting point, the following analogy between science
 289 and network industries is proposed. Let’s say a scientific discipline is like a table to which
 290 scientists add contributions, and the probability that a given scientist will add to cluster *x*
 291 depends on the share of *x* in the discipline. This results in increasing returns to adoption
 292 because the more people join a cluster, the bigger the odds that a newly born scientist will
 293 join that cluster. An idea by Alvin Goldman helps to make this step intuitive. Upon entry,
 294 the newborn scientist finds himself in what Goldman called the novice/expert-problem.

295 The novice/2-experts problem is whether a layperson can justifiably choose one putative
296 expert as more credible or trustworthy than the other with respect to the question at hand, and
297 what might be the epistemic basis for such a choice? (Goldman 2001, p. 92)

298 I maintain that the novice is incapable of making his choice on the basis of scientific
299 content. By definition, the novice has not yet spent any energy in getting to know the field,
300 so he cannot rely on his knowledge of the different clusters, he cannot assess the reliability
301 of different experts and he has no perfect knowledge of the state of the discipline which
302 would enable him to compare the available evidence for different clusters or to know
303 which network has the most adopters. Instead, what I propose is simply that the probability
304 that the novice contributes to network x is equal to the share of x in the discipline. This can
305 be interpreted as the idea that the novice is likely to join the cluster to which his peers
306 adopt. And the odds that those peers will be members of network x is equal to the share of
307 x in the discipline. So what is taken out of the equation here is that we require agents to be
308 perfectly informed about the state of the discipline or have some prior knowledge about its
309 contents (which would make novices not so novice after all).

310 Once a scholar is in the field, he will get to know the available evidence and arguments.
311 This will reinforce the mechanism that larger clusters increase the odds of a scientist
312 joining, because the more adopters a cluster has, the better the quality will be of the
313 arguments and evidence produced in it. Moreover, as a scholar stays longer in the cluster,
314 sunk costs increase, making change of network more costly. So in the model proposed, a
315 scientist shifting between networks is possible but becomes increasingly unlikely as the
316 scientist ages.

317 Then what does it mean to join a network? Each network consists of nodes
318 (contributions) that are held together by a certain *standard*. The only contributions that can
319 be added to the network are those compatible with that standard. As for scientific
320 communities, a standard specifies which basic concepts to use, what theoretical
321 assumptions to make, what the important problems are and what counts as a good solution.
322 Note that the way a standard in science is characterised here is analogous to concepts such
323 as ‘paradigm’, ‘research programme’, ‘schools of thought’, etc. The analogy between
324 science and network industries then means that a scientist working within a certain
325 paradigm can be approached in the same way as a computer user adapting to an operating
326 system (e.g. Apple, Windows).

327 An account based on increasing returns means that science is driven by compatibility.
328 Why are scientists more likely to adapt to bigger networks? Because there are *network*
329 *externalities*. The strength of the network rises exponentially with the number of adopters.
330 The larger the network a scholar contributes to, the more people he will reach, the more
331 impact it will have and the less costly his effort will have been. The bigger the network, the
332 more gains from network externalities and the more attractive it is to join it. This makes
333 sense, at the aggregate level, the intrapersonal level and the personal level. At the
334 aggregate level, scientists are involved in a never-ending quest to increase the size of the
335 network they adapt to,⁸ refining their own network to resist acquisitions by others and
336 discussing with others in order to find common ground and spot potential synergies. From
337 this point of view, scientific research can be seen as performing due diligence; unification
338 and ‘economics imperialism’ can then respectively be seen as mergers and acquisitions.
339 This drive for compatibility provides an explanation for ‘unification’ as an ideal in science.
340 At the interpersonal level, the drive for compatibility explains why scientists try to
341 persuade each other. In a world like Kitcher’s, there is a premium on having as little as
342 possible other scientists pursuing the same path as yourself, because every new scientist
343 means a new competitor (i.e. decreasing returns to adoption). Scientists travelling the

344 world to share their ideas and making efforts to persuade one another is a truly puzzling
 345 occurrence from this point of view. It does not make sense in an environment with
 346 decreasing returns to adoption. If a scientist believes he is on the right track, the last thing
 347 he will want to do is to inform others, because this increases the other's chance of winning
 348 the prize and decreases his. The assumption of increasing returns to adoption, however,
 349 can account for this otherwise irrational behaviour. Thirdly, on a psychological level, the
 350 drive for compatibility seems to be hard-wired in the human brain since people tend to
 351 avoid cognitive dissonance.

352 What can be said about the epistemic quality of the result of this process? At the
 353 individual level, scientists try to reap the benefits of network externalities. At the
 354 aggregate level, each network produces its set of coherent, empirically adequate evidence;
 355 however, the networks are more or less incompatible (or, in more philosophical
 356 terminology, incommensurable) in the same sense as Microsoft and Macintosh. On the one
 357 hand, it is understandable why incommensurability elicits such horror (for it implies
 358 serious limits to compatibility). However, a network perspective also injects this prospect
 359 with some hope, because, as Gottinger notes about network industries: 'compatibility
 360 should be conceptualized as a continuum' (Gottinger 2003, p. 5). In networks, 'converters'
 361 can be developed between networks, e.g. a program to run Microsoft software on a
 362 Macintosh computer. These converters (cf. bridge principles) can work both ways, as
 363 could be said to be the case when unification in science takes place, but it might as well
 364 only work one way (as was the case with running Microsoft software on a Macintosh
 365 computer). In science, the equivalent of the development of a one-way converter might be
 366 the discovery of theoretical tools to reduce one body of theory to another.

367 How should different networks then be evaluated? Network industries tend to develop
 368 a dominant standard (or, applied to science, a dominant paradigm). What is its status?
 369 Scientific realism would hold that a successful cluster of ideas, successful in the sense of
 370 being coherent, empirically adequate and popular among scientists, is most likely to be
 371 closest to the truth. However, a model exhibiting positive feedback mechanisms like the
 372 Arthur model implies that a specialisation bias is inherent to its dynamics. In a system
 373 driven by a quest for compatibility, the system as a whole is likely to bring up a dominant
 374 standard without the necessity to assume that this is the result of any truth-tracking ability
 375 possessed by the dominant standard. Arthur's paradigm model described in section 3
 376 yields a dominant technology every time it is played. However, even with fixed initial
 377 conditions, the technology dominating varies every single time the game is played,
 378 indicating that the popularity of the technology does not depend on any inherent qualities
 379 of the technology itself but rather is brought about by network externalities. As such,
 380 within this perspective success is no proxy for truth.

381 382 383 **5 Economics as a network industry**

384 By way of illustration, let me apply this view to the discipline of economics. Although
 385 there is an obvious circularity in assessing the economics discipline through a perspective
 386 based on one of its own models (cf. Mäki 2005, p. 218), I do believe it might offer a fruitful
 387 alternative perspective to counteract the influence of models assuming an invisible hand
 388 (and often, as a consequence, decreasing returns) without even being explicit about it.

389 Science as a network industry means that there is a significant specialisation bias. Can
 390 any signs of this be detected in economics? The dominance of 'neoclassical economics'
 391 has been lamented by so-called 'heterodox economists' and pleas for more pluralism in
 392 economics are widespread.⁹ However, Colander Holt, and Rosser (2004) warn against

393 exaggerating the homogeneity of dominant strands in economics and emphasise the
 394 changes the discipline has gone through since the 1960s, away from what they call ‘the
 395 holy trinity – rationality, selfishness, and equilibrium –’ (Colander et al. 2004, p. 485).
 396 Sent (2006) even goes one step further and argues that economics has failed to become a
 397 unified discipline, despite several serious attempts.

398 However, there seems to be a rather widespread discontent with how economics is
 399 faring as a discipline, giving rise to e.g. the Post-Autistic Economics movement and
 400 economists calling themselves ‘heterodox’. One of its main proponents, Tony Lawson,
 401 acknowledges the argument by Colander et al., but adds:

402 But it remains the case that these and all other widely sanctioned examples of ongoing change,
 403 diversity, novelty, complexity, evolution and multi-dimensionality, etc., are occurring within
 404 the framework of formalistic modelling. The insistence on *mathematical–deductive*
 405 modelling prevails in all cases; the essential feature of the recent and current mainstream
 406 remains intact. (Lawson 2006, p. 491; emphasis added)

407 On this, Colander et al. seem to agree with Lawson:

408 Our view is that the current elite are relatively open minded when it comes to new ideas, but
 409 quite closed minded when it comes to alternative methodologies. If it isn’t modeled, it isn’t
 410 **Q3** economics, no matter how insightful. (Colander et al. 2004, p. 493)

411 Numerous authors have characterised the above view as a caricature of the discipline,
 412 but also the less outspoken variants point to an interesting puzzle: how can the relatively
 413 high rate of homogeneity in economic theory be reconciled with the low rate of decisive
 414 empirical information, to wit information that would provide substantive reasons to prefer
 415 certain approaches over other less empirically successful ones? The view of scientific
 416 activity as a network industry proposed in this paper suggests that this situation can be
 417 interpreted as follows: (parts of) the economics discipline has (been) locked into a
 418 ‘mathematical-deductive modeling’ standard. This would explain the repeated pleas for
 419 pluralism and is in line with the existence of shared basic concepts and theoretical
 420 assumptions within large parts of economics. The presence of a broadly accepted standard
 421 allows for extensive division of labour and hence results in strong specialisation within the
 422 standard and large differences in appeal compared to rival approaches based on less
 423 popular standards. The broad acceptance of a certain standard in economics can explain
 424 why economics textbooks tend to be a-historic, giving economics students little or no
 425 knowledge of rival approaches, while textbooks in other social sciences typically do
 426 contain information on the development of the discipline and possible alternative
 427 standards. Because of its large rate of adoption, the standard itself can be taken for granted
 428 and needs no explicit legitimation anymore. Given that the power of a network crucially
 429 depends on its size (it rises quasi-exponentially with the number of adopters), the broad
 430 acceptance of a certain way of doing economics gives rise to strongly integrated networks
 431 in which non-conformists are sanctioned.

432 to get an article published in most of today’s top rank economic journals, you must provide a
 433 mathematical model, even if it adds nothing to your verbal analysis. I have been at seminars
 434 where the presenter was asked after a few minutes, ‘Where is your model?’. When he
 435 answered ‘I have not got one as I do not need one, or cannot yet develop one, to consider my
 436 problem’ the response was to turn off and figuratively, if not literally, to walk out. (Lipse
 437 2001, p. 184)

438 Simultaneously, the dominant standard is characterised by an abundance – almost to the
 439 point of redundancy – of slightly different contributions.
 440
 441

If design and manufacture have fixed costs, then there are too few products designed for networks or standards with few users, and too many for those with many users. This reinforces the winner-take-all characteristic of these markets. (Heal 1998, p. 12)

6 Scientific pluralism

Following the illustration of economics as a network industry in the previous section, it is now time for more general conclusions. How do increasing returns affect the tradeoff between specialisation and diversity? And what are the implications for pluralism? Pluralism involves a normative endorsement of diversity of views and as such it formulates an answer to what position on the tradeoff *should* be pursued. A tentative answer to this question can now be formulated.

The impact of network externalities on network industries produces a specialisation bias. By this I mean that a group of scholars (*ceteris paribus*, for example institutions which constitute the group have to be strong enough to support large groups) will be driven by the same forces that gives for example software companies a natural tendency towards lock-in/monopoly/winner-take-all. It provides scientists with an incentive to follow those strands which are already well established and to keep away from marginal research programmes. In other words, from this point of view the extra-scientific factors which Kitcher (1990) refers to¹⁰ to avoid the CO–IR discrepancy are not enough to counter the effects of network externalities on scientific communities.

One of the lessons of Kitcher (1990) is that the impact of social factors on the division of labour in the sciences needn't be disruptive for the achievement of our epistemic goals, but might indeed be essential for it. If scientists were all rational, they might all choose the same research path, consisting of the one which has, based on the available evidence, the highest chance of success. Thanks to lesser motives such as the thirst for fame and fortune, scientists divide their labour across different research paths because a path's lower chance of success is compensated by the smaller amount of competitors. As such, his idea of what *should* be done becomes clear in this passage:

The very factors that are frequently thought of as interfering with the rational pursuit of science – the thirst for fame and fortune, for example – might actually play a constructive role in our community epistemic projects, enabling us, as a group, to do far better than we would have done had we behaved like independent epistemically rational individuals. Or, to draw the moral a bit differently, social institutions within science might take advantage of our personal foibles to channel our efforts toward community goals rather than toward the epistemic ends that we might set for ourselves as individuals. (Kitcher 1990, p. 16)

In Kitcher's model extra-scientific factors cause diversity and this diversity is then efficiently allocated by an invisible hand. Pluralism seems to be a matter of *laissez-faire*, with institutions merely to make sure this invisible hand can operate. On the other hand, a model incorporating increasing returns will follow a different dynamics, leading to different implications for pluralism and opposing *laissez-faire*.

To compare Kitcher's pluralism to the one proposed in this paper, I will position them in the classification developed by Van Bouwel (in press) which distinguishes three different views on scientific pluralism. Each one of them offers a different answer to the question what a good balance between specialisation and diversity should be and how to achieve that balance.

Consensual pluralism relies on the invisible hand to sort out different views. As such, the tension between specialisation and diversity disappears. No dissent is needed. Its attitude toward the tradeoff is one of *laissez-faire*. Groups will self-organise toward their optimal balance. An example is Kitcher (1990), claiming that diversity is necessary to

491 avoid a discrepancy between individual rationality and the collective optimum. The
 492 related answer to what should be done is then a laissez-faire approach. Philosophers of
 493 science in *turn* have – implicitly or explicitly – made use of the idea of the invisible hand
 494 to explain how diversity and dissent can still be consistent with rationality and objectivity
 495 in science.¹¹ This leads to pluralism without dissent.

496 *Agonistic pluralism* stresses the importance of dissent instead of relying on an invisible
 497 *hand*. It suggests rules for active engagement such as those provided by Helen Longino.
 498 The focus is not on finding the right balance between specialisation and diversity, but on
 499 creating an adequate environment for fair discussion. Which concessions can be made
 500 with respect to the tradeoff will be negotiated between communities of inquirers.

501 *Antagonistic pluralism* prefers to shift the balance toward specialisation, at the cost of
 502 diversity. For example Kuhn sees innovation and revolution not as products of diversity,
 503 but as the result of sustained specialisation.¹² As such he seems to be in favour of shifting
 504 the balance towards more specialisation:

505 history strongly suggests that, though one can practice science [. . .] without a firm consensus,
 506 this more flexible practice will not produce the pattern of rapid consequential scientific
 507 advance [in which] development occurs from one consensus to another, and alternate
 508 approaches are not ordinarily in competition. (Kuhn 1977, p. 232)

509 Jouni-Matti Kuukkanen (2007) acknowledges that, especially in his later work, Kuhn
 510 viewed the development of science as a strong tendency toward internal coherence
 511 resulting in a plurality as a sum of exclusive monisms; a fragmentation of science.¹³ As
 512 such, antagonistic pluralism is the only standpoint which has an actual preference for a
 513 certain position on the tradeoff, whereas consensual pluralism opts for laissez-faire and
 514 agonistic pluralism sees it as a matter of negotiation.

515 The three views can be schematised as in Table 1.

516 While Kitcher's pluralism falls in the first category, the pluralism defended in this
 517 paper falls into the third. But unlike Kuhn, the reasons to support this view are not
 518 historical but come as a result of the view of science as a network industry. On this view,
 519 specialisation is not so much a desideratum but a consequence of the assumption of
 520 increasing returns itself. As such, there is no other option than to accept that science is
 521 driven by an urge to specialise and a quest for compatibility. This means that I see no other
 522 option than to defend an antagonistic pluralism, in which different, more or less
 523 incompatible clusters of basic concepts and key theoretical assumptions strive for
 524 dominance over a discipline, although this dominance is not stable and no guarantee for
 525 superiority (respectively accounting for the existence of scientific revolutions and the
 526 pessimistic meta-induction). The reason for this acquiescence with the specialisation bias
 527 of science is that it produces enormous network benefits which would be lost with more
 528 diversity. However, full specialisation should be avoided, for the same reasons as
 529 governments enact antitrust policy in economic sectors where increasing returns play a
 530 large role (cf. Microsoft vs. EU).

531 This view responds to consensual pluralism by opposing a laissez-faire attitude.
 532 Systems exhibiting increasing returns do not tend to find an optimal equilibrium; rather

533
 534
 535 **Q10** Table 1.

536 Consensual	invisible hand	laissez-faire	'epistemic liberalism'
537 Agonistic	discussion	engagement/dissent	'epistemic democracy'
538 Antagonistic	specialisation	dominance dissidence/indifference	'epistemic hegemony'

they have the tendency to lock into suboptimal states. This means that arguments for laissez-faire, letting the ‘market for ideas’ sort itself out, are highly problematic in the context of network industries.

The long-standing public policy concerns over network industries are not accidental, because those industries often embody two major and widely recognized forms of potential market failure: significant economies of scale – with the potential for monopoly – and externalities. (White 1999, p. 1)

In response to agonistic pluralism, the present view acknowledges the importance of dissent. Dissent is important in identifying possibilities for merger and for finding powerful gateways. However, dissent and discussion are already hard-wired into the idea of scientific research as due diligence and need no further encouragement from an institutional design perspective. Rather than stimulating discussion between different networks, the present view acknowledges the (albeit not total) differences between networks, their potential for specialisation and their singularity.

In the end, this version of antagonistic pluralism comes surprisingly close to democracy because it resembles the same structure: a few large competitors each having their own monist worldview, but limited in their claims to hegemony by democratic institutional structures (without which a lock-in would probably emerge, such as dictatorship).

7 The scope for institutional design

The previous section was concerned with how specialisation and diversity *should* be traded off. But ought implies can. *How* can a group’s balance between specialisation and diversity be changed? Who or what decides? How can a good position be reached, or does it come by itself? From the perspective of science as a network industry a number of policy recommendations and other possibilities of action can be formulated.

Perhaps disconcertingly, the prospects for science policy in a context of increasing returns must remain modest because of the unpredictability of the measures taken.

To the extent that small events determining the overall path always remain beneath the resolution of the economists’ lens, accurate forecasting of an economy’s future may be theoretically, not just practically, impossible. (Arthur 1994, p. 12)

However, according to Arthur, possibilities remain for policymakers:

Steering an economy with positive feedbacks into the best of its many possible equilibrium states requires good fortune and good timing – a feel for the moments when beneficial change from one pattern to another is most possible. Theory can help identify these states and times, and it can guide policymakers in applying the right amount of effort (not too little but not too much) to dislodge locked-in structures. (ibid., p. 12)

The general challenge in a context of increasing returns seems to be to reap the benefits of specialisation while not eliminating diversity altogether. In fact, because of the presence of network industries in the economy, governments have already developed extensive regulation to do just that: anti-trust policy. ‘The essence of anti-trust policy is to find ways of keeping these economies of scale or of standardisation while bringing in competition’ (Heal 1998, p. 4). The implementation of anti-trust law is seen as self-evident in most market economies, so why not in science if science is seen as a network industry? The main goal should then be to avoid groups from locking in or at least locking in to suboptimal standards. For science, this could be achieved by increasing the possibility of scientists shifting between networks. One way to do this is to promote the creation of (what in network theory is called) gateways: ‘gateways [...] create the possibility to have

different types of (incompatible) networks communicate and work together' (Gottinger 2003, pp. 4–5). However, the creation of bridge principles, reduction and unification seem to fall beyond the scope of science policy and are more a matter of genuine scientific labour. Indeed, science as due diligence implies that scientists are already on the case. A more feasible measure to reduce the odds of a lock-in could be to increase scientists' knowledge of alternatives. Returning to the case of economics discussed above, it is striking that history of economics has all but disappeared from economic curricula worldwide. Not only might this have contributed to the apparent lock-in in which it appears to find itself today, but also it might suggest a remedy. Greater awareness of the history of economic thought could increase liquidity in the economics discipline. Perhaps surprisingly, teaching the history of economic thought might thus turn out to be the academic equivalent of free trade. A second way to achieve this is to try and assuage the specialisation bias. For economics, but also for science in general, this could be done for example by reducing the importance attributed to bibliometric methods. As illustrated with an example in section 4, bibliometric methods increase the specialisation bias. Academic success is increasingly being defined in terms of bibliometric criteria. There is an ever-increasing incentive to publish in highly cited journals, which puts an extra premium on compatibility.

8 Conclusion

Economics is a typical resource for social epistemology and the division of labour is a common theme for economics. As such it should come as no surprise that the present paper has turned to economics to formulate a view on the dynamics of scientific communities, with precursors such as Kitcher (1990), Goldman and Shaked (1991) and Hull (1988). But although the approach is similar to theirs, the view defended in this paper was different. The aforementioned contributions assume decreasing returns to adoption. Mäki (2005) points out that the lessons we can draw from economics can often go either way depending on the model one chooses. Thus, the aims of this paper have been (1) to illustrate this flexibility by putting forth a model assuming increasing returns to adoption in science rather than decreasing returns, viewing scientific activity as a network industry; and (2) to outline the implications of this view for scientific pluralism and institutional design.

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Notes

1. 'Each person, therefore, making a tenth part of the forty-eight thousand pins, might be considered as making four thousand eight hundred pins in a day. But if they all had wrought separately and independently, and without any of them having been educated to this particular business, they certainly could not each of them have made twenty, perhaps not one pin in a day' (Smith 1977 [1776]).
2. 'That is why I speak of an "essential tension" implicit in scientific research. [...] Very often the successful scientist must simultaneously display the characteristics of the traditionalist and of the iconoclast' (Kuhn 1977, p. 227; emphasis added).
3. 'I shall call the advice to select from a number of theories the one that promises to lead to the most fruitful results, and to stick to this one theory even if the actual difficulties it encounters are considerable, the *principle of tenacity*. [...] if change of paradigms is our aim, [...] we must be prepared to accept a *principle of proliferation*. [...] the interplay between tenacity and

- 638 proliferation [...] is [...] an essential feature of the actual development of science’
 639 (Feyerabend 1970, pp. 203–209; emphasis added).
- 640 4. ‘[S]tudents of the development of science, whether sociologists or philosophers, have
 641 alternately been preoccupied with explaining *consensus* in science or with highlighting
 642 disagreement and *divergence*. [...] neither approach has shown itself to have the explanatory
 643 resources to deal with both’ (Laudan 1984, p. 2; emphasis added).
 - 644 5. This tension is noted in for example Heal (1998) and discussed at length in Warsh (2006).
 - 645 6. The main argument for this inconsistency has been provided by Coase (1937).
 - 646 7. Blinder, Canetti, Lebow, and Rudd (1998) estimate that 89% of firms are subject to constant or
 647 falling marginal costs.
 - 648 8. It is the network that determines what the relevant questions are and what count as good
 649 solutions. Within a network, scientists will most likely be competitors, e.g. to be the first to
 650 solve a problem and win a prize. But at the network level, they have a shared interest in the
 651 expansion of the network relative to other networks because that means more people find the
 652 problem important.
 - 653 9. E.g. ‘Plea for a Pluralistic and Rigorous Economics’, an advertisement in *American Economic
 654 Review*, calling for ‘a new spirit of pluralism in economics, involving critical conversation and
 655 tolerant communication among different approaches’.
 - 656 10. ‘The very factors that are frequently thought of as interfering with the rational pursuit of
 657 science – the thirst for fame and fortune, for example – might actually play a constructive role
 658 in our community epistemic projects, enabling us, as a group, to do far better than we would
 659 have done had we behaved like independent epistemically rational individuals’ (Kitcher 1990,
 660 p. 16).
 - 661 11. This point is made by Ylikoski (1995, p. 35).
 - 662 12. ‘At least for the scientific community as a whole, work within a well-defined and deeply
 663 ingrained tradition seems more productive of tradition-shattering novelties than work in which
 664 no similarly convergent standards are involved. How can this be so? I think it is because no
 665 other sort of work is nearly so well suited to isolate for continued and concentrated attention
 666 those loci of trouble or causes of crisis upon whose recognition the most fundamental advances
 667 in basic science depend’ (Kuhn 1977, p. 234).
 - 668 13. ‘Especially in his later writings, Kuhn argued more or less directly that scientists have
 669 attempted to increase coherence in the history of science. However, Kuhn did not make a claim
 670 that the total coherence of science has increased in the course of history, and moreover, it is
 671 hard to find any historical evidence for it. Kuhn held that scientists’ attempts to improve the
 672 coherence of their theories paradoxically tend to decrease the total coherence of science,
 673 because that activity leads to the fragmentation of science. This kind of speciation or
 674 specialisation of scientific fields is how Kuhn came to understand scientific revolutions in the
 675 latter part of his career’ (Kuukkanen 2007, p. 556).

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