

How To Have a Career in Pasteur's Quadrant

[Excerpted from *Ectotherms in Changing Environments*.

Working in Pasteur's Quadrant, unpublished manuscript.

Supported in part by NSF grant 14- 51931 "OPUS: Ectotherms
in Changing Environments".]

October 23, 2020

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Context

I closed the book with a chapter on career advice, in particular how to have a career in which one is motivated by important applied problems and simultaneously seeks fundamental understanding. Donald Stokes (Stokes, 1997) called this Pasteur's Quadrant (explained below). The editor at Princeton University Press suggested that even if Princeton were to publish the book, I remove this chapter and publish it as a stand alone paper; this is what got me thinking about making all the book into stand alone papers rather than trying to rewrite. For obvious reasons, I have written the chapter with ecologists in mind, and the references are heavy with applications of State Dependent Life History Theory (SDLHT) implemented by Stochastic Dynamic Programming (SDP).

The topics covered are these:

- An Epigram from Phil Levin.

- 25 ● Introduction (a description of Pasteur’s Quadrant and examples of the use of SDLHT
26 in Pasteur’s Quadrant).
- 27 ● Working Across Disciplinary Boundaries.
- 28 ● Professional Preparation (undergraduate, graduate, post-doctoral studies; working
29 hours and work life balance; finding mentors at every point your career).
- 30 ● Communicating Science (writing and speaking from the short talk to the job inter-
31 view; using practice to improve your speaking skills).
- 32 ● The Far Beyond (The impostor syndrome, the importance of stupidity in science, and
33 post-decision regret; academic or non-academic positions and leaving and returning
34 to academia, becoming part of a (not too big) research group, and ethical issues in
35 funding).

36 **Epigram from Phil Levin**

37 Phil Levin's reconstruction of a poster he did on Louis Pasteur in elementary school
38 (ca.1972; see <http://www.washingtonnature.org/fieldnotes/2017-science-what-inspires-me>):

39
40

Louis Pasteur discovered that the growth of microbes was responsible for spoiling milk

And

he invented a process of heating milk to kill the microbes and prevent it from spoiling.

Louis Pasteur discovered that a disease caused by microbes was killing silkworms

And

41 *he developed a process to eliminate the microbes, protect the silkworms and ultimately save the French silk business.*

Louis Pasteur discovered that he could artificially weaken disease-causing microbes

And

he created some of the earliest vaccines, exposing people to the weakened microbe to foster immunity to the stronger form.

42 **Introduction**

43 Levin continued "It was the 'ands' that stuck with me. Some of Pasteur's peers were solely
44 on a quest for fundamental understanding. Others, were less interested in knowledge for
45 knowledge's sake – but rather sought to practically apply existing knowledge. Pasteur was
46 different. He pushed the frontiers of knowledge, but did so because he saw a real-world
47 need. Pasteur sought to uncover nature's secrets - and use this wisdom to improve the
48 human condition...We conservation scientists must follow in Pasteur's footsteps – steadfast
49 in our learning and resolute in its application. By conducting such use-inspired research,

50 our scientists have the best hope of developing novel, practical, applicable and scalable
51 solutions to the wicked problems our ecosystems face.” Most of this chapter is about how
52 (rather than why) to have a career in Pasteur’s Quadrant.

53 Ludwig et al (2001) address the why: “More than any previous generation, the current
54 generation of ecologists is concerned about applications of their work to society’s problems.
55 Since the first Earth Day about 30 years ago there has been a change in the public
56 perception of environmental problems and a corresponding growth in the sensitivity of
57 ecologists to issues of public policy” and reiterated the question asked by Roszak in 1972:
58 “Which will ecology be: the last of the old sciences or the first of the new?” (Roszak
59 1972, pg 404). We’d like the answer to be that ecology is the first of the new sciences.

60 Since 1980 – that is, since I began my academic career – ecologists and evolutionary
61 biologists working in universities slowly but inexorably shifted from disdaining work that
62 had application to seeking it. In part this is due to changed funding opportunities, but it
63 is also due to the demands of our students that work be both scientifically first rate and
64 important.

65 Simply put: we must do science that matters and science that matters will be mo-
66 tivated by an important applied problem and will cross disciplinary boundaries. Today,
67 of course, everyone – including funders – is talking about work that spans disciplinary
68 boundaries; indeed *Nature* had an entire special feature on interdisciplinary work in 2015
69 (e.g. Brown et al 2015, Ledord 2015, Pettitt 2015, Rylance 2015, van Noorden 2015).

70

71 **Pasteur’s Quadrant**

72 In his book *Pastuer’s Quadrant* Donald Stokes (1997) argued that a single axis between
73 basic and applied science is the wrong way to think about things. Rather, one must
74 focus attention in a plane. One axis of assessment of work is whether it is motivated
75 by a consideration of use and the other is whether there is a quest for fundamental
76 understanding. The plane can then be divided into the four quadrants (Figure 1) 1) No
77 consideration of use and quest for fundamental understanding; 2) Consideration of use
78 and a quest for fundamental understanding; 3) Consideration of use and no quest for
79 fundamental understanding; and 4) No consideration of use and no quest for fundamental
80 understanding.

81 Stokes called the quadrant ‘not motivated by consideration of use and search for funda-

| | | Considerations of use? | |
|--------------------------------------|-----|-------------------------------|--|
| | | No | Yes |
| Quest for fundamental understanding? | Yes | Pure basic research (Bohr) | Use-inspired basic research (Pasteur) |
| | No | | Pure applied research (Edison) |

Figure 1: Stokes's (1997) vision of science is that the dichotomy between basic and applied science is a false one. Rather in every scientific endeavor we may ask about application – is there consideration of use motivating this work – and whether or not there is a quest for fundamental understanding. Niels Bohr provides the canonical example of an individual whose work was not motivated by consideration of use but involved the deep search for fundamental understanding and Thomas Edison one whose work was motivated by consideration of use but did not search for fundamental understanding. Louis Pasteur – whose work from the time of his PhD was motivated by an important applied problem (Debré 1994) while he simultaneously sought fundamental understanding in his work.

82 mental understanding' Bohr's Quadrant, the one 'motivated by consideration of use and
83 no search for fundamental understanding' Edison's Quadrant and the quadrant 'motivated
84 by consideration of use and search for fundamental understanding' Pasteur's Quadrant
85 since Pasteur's entire career involved work that was motivated by an important applied
86 problem but simultaneously sought fundamental understanding (Debré 1994). Pasteur
87 himself said "There is no such thing as a special category of science called applied science;
88 there is science and there are its applications, which are related to one another as the
89 fruit is related to the tree that has borne it (Debré 1994)".

90 There are already many examples of the use of SDP models for work in Pasteur's
91 Quadrant. In Table 10.1, I show a smattering of papers more or less randomly chosen to
92 illustrate some of the applications (also see Chapters 7 (on discarding in fisheries), 8 (on
93 conservation biology), and 9 (on agroecology) in Clark and Mangel (2000)).

94 **Table 10.1. Some Examples of the Use of State Dependent Behavioral or Life**
 95 **History Theory in Pasteur’s Quadrant Since 2000**

96

| <u>Authors</u> | <u>What They Did</u> |
|--------------------------------|---|
| Babcock and Pikitch (2000) | Modeled the targeting decisions of bottom trawlers to understand the consequences of management actions and market forces on the fishery and the fish. |
| Shea and Possingham (2000) | Used in SDP in a metapopulation context to determine optimal number and size of releases of biological control agents. |
| Westphal et al (2003) | Used a SDP model to determine optimal restoration policies for the metapopulation of an endangered Australian bird. |
| 97 Costello and Polasky (2004) | Developed a SDP for choosing sites to include in a network of marine reserves, accounting for the inability to protect all sites immediately, and the benefits to biodiversity from different sites |
| Meir et al (2004) | Used SDP to assess biodiversity consequences when a network of reserves cannot be implemented all at once but rather over time and derived simple rules of thumb. |
| Rout et al (2007) | Used a SDP model to determine optimal translocation strategies for threatened species, comparing minimizing long-term loss and maximizing short-term gain. |

98

99

100 **Table 10.1 (continued) Some Examples of the Use of State Dependent Behav-**
 101 **ioral or Life History Theory in Pasteur’s Quadrant Since 2000**

102

| <u>Authors</u> | <u>What They Did</u> |
|----------------------------|--|
| Bogich and Shea (2008) | Used SDP models to determine the most efficient management strategy for gypsy moth (<i>Lymantria dispar</i>) and developed a rule of thumb that, depending on the state of the system, the optimal action was eradicating medium-density infestations, reducing large-density infestation or reducing the colonization rate from the main infestation. |
| Bauer et al (2010) | Explored potential explanations for observations that Red Knots use a diverse array of migration routes, with potential application to habitat protection. |
| Baruch-Mordo et al (2013) | Used the basic patch selection model of Mangel and Clark (1988) to develop a decision support tool to evaluate mitigation strategies in human-wildlife conflict. |
| Hackett and Bonsall (2016) | Used SDP to determine optimal rotation strategies between treatments and refuges to delay development of resistant in <i>Bacillus thuringiensis</i> (Bt) treated crops. |

104

105 **Table 10.1 (concluded) Some Examples of the Use of State Dependent Behav-**
106 **ioral or Life History Theory in Pasteur’s Quadrant Since 2000**

107

| <u>Authors</u> | <u>What They Did</u> |
|-----------------------------|--|
| Kling et al (2017) | Using SDP with information as a state variable in a bioeconomic model of lionfish invasion and control. |
| Visscher and Merrill (2018) | Used SDP to explore how interpatch distances on a landscape affected ruminants moving across that landscape. |

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110

111 This chapter is a ‘lessons learned’ from my career for young scientists who want to work
112 in Pasteur’s Quadrant (also see Mangel 2017¹). To do so, we will first discuss work that
113 crosses disciplinary boundaries, and then turn to the relevant professional preparation.
114 In that section, we will go forward in time – considering undergraduate school, graduate
115 school, postdoctoral work, and beyond (including both academic and nonacademic work).
116 We will then turn to communicating science. We close this chapter with advice from
117 Pasteur himself.

118 **Working Across Disciplinary Boundaries**

119 The words “interdisciplinarity” and “transdisciplinarity” are quite the rage, but there are
120 problems for both defining the kind of work and achieving successful outcomes (Ledford
121 2015). For example, does interdisciplinary mean crossing between physiology and behav-
122 ioral ecology? I believe that interdisciplinary should mean something more than crossing
123 silos in the biological science.

124 In the recent book edited by Phil Levin and Melissa Poe (Levin and Poe 2017) on
125 marine conservation in the anthropocene many chapters use the word transdisciplinary.
126 For example, Arkema and Ruckelshaus (2017) wrote: “Transdisciplinarity is the youngest
127 of the trio of terms that also includes multidisciplinary and interdisciplinarity (Choi and
128 Pak, 2006). Although definitions differ across fields, in general, multidisciplinary research

¹This paper is one of a series in the “Food for Thought” section of the *ICES Journal on Marine Science* in which senior scientists write brief autobiographies in which they look back and provide advice for young scientists. They are open access and can be found at <https://academic.oup.com/icesjms/pages/luminaries>

129 draws on knowledge from different disciplines but stays within individual disciplinary
130 boundaries. Interdisciplinary research links two or more disciplines by integrating theo-
131 retical frameworks, methodology, and the perspectives and skills from several disciplines.
132 Transdisciplinary research creates new conceptual, theoretical, methodological, and trans-
133 lational innovations to societal and related scientific problems concurrently by integrating
134 various societal and scientific bodies of knowledge”. I have provided their citation to Choi
135 and Pak (2006), and another (Lang et al 2012) for your convenience.

136 Lang et al (2012) focus on sustainability science, rather than medical science as in
137 Choi and Pak (2006) They describe three phases of the work.

138 • *Phase A: Collaborative work to frame the problem and build the research team.* This
139 phase includes identifying and describing the problem to be solved, agreeing on the
140 research objectives, which includes collaborative formulation of the strategy (broad
141 objectives) and tactics (specific research questions), and designing the conceptual
142 framework to be used and ways that knowledge across disciplines will be integrated.
143 The key here is to find the match between complexity of the problem and the science
144 that is to be done. No science – experimental, field, or theoretical – can ever be
145 ‘realistic’, by the very nature of science. See the Appendix to Chapter 2 of Mangel
146 and Clark (1988) or Chapter 4 in Clark and Mangel (2000) about why. It is not
147 unsurprising, but good to know, that collaborative development of a model increases
148 the trust in it and its use (Ulibarri 2018).

149 • *Phase B: Using collaborative work to create new and transferrable knowledge.* This is
150 when the research gets done, which requires agreeing on a set of methods, developing
151 them as needed, and applying them to the problem. If the collaboration is effective,
152 different disciplinary skills will be brought to bear on the focal problem. It is key here
153 to situate the problem and to identify the particular contributions to the solution
154 of the problem, and how the disciplinary approaches will be integrated. The earlier
155 one involves stakeholders in the process, the more likely one is to get buy-in on their
156 part.

157 For example, Benson and Stephenson (2017) argue that interdisciplinary work in
158 fisheries is done in the space of scale of the problem, nature of the application, and
159 the type of advice required for fisheries management. Current approaches provide
160 tactical, prescriptive advice at the scale of fishery and address a small fraction of
161 the decision space required for holistic management. The types of questions that

162 need to be addressed include i) *Management Context*: Here one asks what are the
163 management objectives and how will one evaluate the success or failure of man-
164 agement. ii) *Inventory the Methods*: Valiela (2001, pg 6) noted “[i]t is a waste of
165 time, of course, to worry about the density of angels on any surface, let alone the
166 head of a pin, unless we have a working seraphometer available.” That is, unless one
167 has an appropriate method – or can develop one – a variety of management objec-
168 tives may remain unattainable. For example, in the case of fisheries, Benson and
169 Stephenson (2017) evaluate the following methods: ecological risk assessment, Man-
170 agement Strategy Evaluation (MSE), multispecies and ecosystem models, dynamic
171 multispecies models, aggregate (biomass production)species models, full ecosystem
172 models that range from Minimum Realistic models (MRM) and Models of Interme-
173 diate Complexity for ecosystem assessments (MICE) through Ecosim with Ecopath
174 to Atlantis, and Bayesian Belief Networks.

- 175 • *Phase C: Integrating and applying the new knowledge*. In this phase, the results
176 of the research are applied and implemented. Lang et al (2012) believe that this
177 needs to be done differently than the usual rubric of ‘from science to practice’. They
178 see it as the integration (or re-integration) of the results into societal practice (e.g.,
179 implementing the evidence based strategies and actions developed during the re-
180 search) and scientific practice (e.g., publishing the results in peer-reviewed scientific
181 journals).

182 Lang et al (2012) consider that in addition to the tangible products of the research (new
183 approaches to the problem, new papers in journals), a properly designed and implemented
184 transdisciplinary process may lead to less tangible products such as high self-efficacy (cf
185 Bandura 1997) of the decision-makers or an intense learning process.

186 Lang et al (2012) also offer design principles. For example in Phase A, one needs to
187 build a collaborative research team that has a joint understanding of the problem to be
188 addressed so that it is possible to define the research project, specific research questions,
189 and decide how one knows when success has occurred. In Phase B, the team leaders must
190 assign the appropriate roles to team members, who then have to adjust and integrate
191 their disciplinary research methods for the problem at hand. In Phase C, one wants to
192 assess the success of the research, develop products for the end-users, and evaluate the
193 overall social and economic impact of the project.

194 Throughout all this, one must be cognizant of the possible modification of objectives

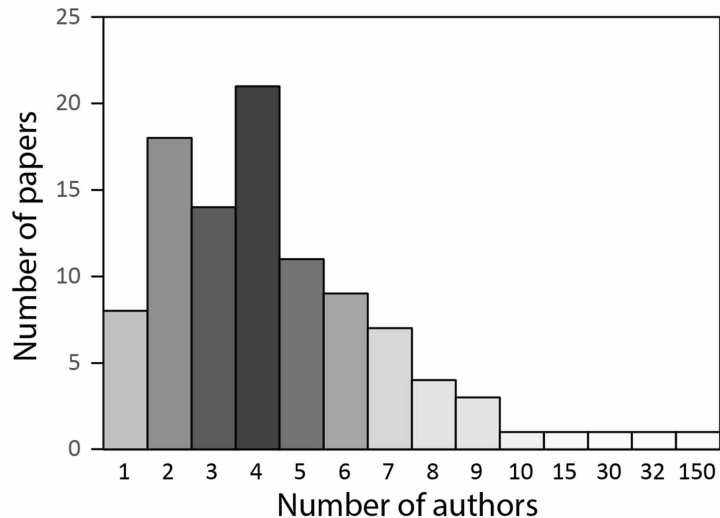


Figure 2: When doing interdisciplinary work, you should plan to become part of a research team Caldera (2016) reviewed multiauthored papers in *Environmental Research Letters* and found that 61 of the top 100 papers had four authors or less. More on the size of research groups below.

195 (Eisenhower once said something like “I have found that in war plans are useless but
 196 planning is essential”), mitigate conflicts between members of the team, and sustain
 197 participation in the project. Those modifications include what you think you can achieve
 198 and how involved you are – learning how to collaborate in interdisciplinary work is more
 199 like a damped spring than a switch. Brown et al (2015) advise researchers (pg 316)
 200 “Build stamina, patience and self-awareness to manage the long journey of establishing a
 201 productive interdisciplinary team. Put your best ideas forward even if they are unfinished,
 202 and be open to alternative perspectives from other disciplines, policymakers, industry
 203 practitioners and community members. Prioritize depth early on, and embrace breadth by
 204 building relationships with those from other fields and practices.” They envision damped
 205 oscillations toward constructive interdisciplinary conversation as a disciplinary scientist
 206 learns to find the golden mean between dominating conversations and passively listening.
 207 Thus, achieving the ability to do interdisciplinary work is not a switch that you turn on
 208 one day, but more like a damped oscillator that overshoots a steady state for some time
 209 but then settles into a regular pattern. Furthermore, you should expect to be part of a
 210 publishing team (Figure 2), but aim for one that is not too large to be effective (below
 211 we will discuss the size of research groups).

212 Palmer et al (2016) describe how the National Socio-Environmental Synthesis Cen-
 213 ter (SESYNC) facilitates interdisciplinary research. Their process includes collaborative

214 project development, project planning, the leads of the interdisciplinary teams having a
215 workshop to plan the project, design and facilitation of teams, and check-in meetings.
216 SESNYC also provides computational support.

217 Hoffmann et al (2017) discuss transdisciplinary integration rather than interdisciplinary
218 work, focusing on the challenges to researchers who want to integrate across disciplines. In
219 this case, the different kinds of knowledge that one can generate, the different participants
220 in the collaboration, and the different levels of involvement by the participants are all
221 crucial. They conclude that interdisciplinary research requires a warm-up stage in which
222 leaders of the project conceptualize what is to be done and what kind of knowledge
223 needs to be integrated, a middle stage in which all the relevant actors are involved in the
224 details of the work, and a final stage of synthesis, which includes the different groups of
225 participants agreeing on the results and communication with the target audience.

226 In my experience, a good estimate for timing is that the first and last stages of the
227 project will take about 10% of the total time and the middle stage will take the remaining
228 80%. Of course, it is worthwhile to remember what my buddy from graduate school, Davis
229 Cope (retired from the Department of Mathematics at North Dakota State University in
230 2017) said he learned about doing a Phd: “One must keep in mind that the first 90% of
231 writing a dissertation takes the first 90% of the time and the last 10% of the dissertation
232 takes the other 90% of the time”. Hoffmann et al (2017) argue that transdisciplinary
233 integration requires strong disciplinary skills for the participants, good interpersonal and
234 management skills of the leaders, and enough time for the integration to occur. I concur
235 with all of these points.

236 Benson and Stephenson (2017, pg 13) found “that two of seven proposed tools to sup-
237 port decision making in the management system can provide tactical advice, but only one
238 (MSE) provides advice that is consistent with our criteria for generation, transmission,
239 and use of scientific information in management advisory processes. MSE ranks similarly
240 well for strategic con- siderations, as it is an approach that meshes tactical, annual de-
241 cision making with longer-term strategic objectives and planning...The EREAF performs
242 similarly well against our criteria as a strategic planning tool. Both MSE and EREAF
243 were developed within the functionalist management science paradigm, which seeks tech-
244 nical (and cost) efficiency in management.. In contrast, the only interpretive management
245 science method that has high applicability in the current management context is BBN.
246 These models help to clarify and define problems using messy, in- consistent, and poten-
247 tially contentious data and objectives...We suggest that BBN could prove very useful as

248 a direction- setting tool for highly complex fisheries problems.”

249 Based on their experiences with water management strategies, Brown et al (2015)
250 suggest that they keys to interdisciplinary work are to

- 251 • *Forge a shared mission* Collaboration requires a common mission between the col-
252 laborators; otherwise one is simply bringing in collaborators for specific technical
253 skills and disciplines are not likely to be crossed in substantial ways A shared mis-
254 sion provides an overall goal (and desired outcome) and must be broad enough to
255 involve all of the disciplinary researchers in a meaningful way. This is really impor-
256 tant because there will be bumps in the road and failures (we need failures in order
257 to learn); shared goals allow researchers to keep the investment of time and effort in
258 the project.
- 259 • *Develop ‘T-shaped’ researchers.* The idea of T-shaped researcher comes from an
260 article in the *Harvard Business Review* (Hansen and von Oetinger 2001). Such
261 researchers are able to succeed in their own discipline but also able to look beyond it.
262 That is, interdisciplinary work requires both breadth and depth (and is more difficult
263 than disciplinary work, for that reason). A T-shaped researcher is highly regarded for
264 contributions in his or her own field of research but also actively engages with other
265 disciplines to understand how they work. In my opinion *to be truly interdisciplinary*
266 *one must both be a serious disciplinary scientist and master the core skills in one or*
267 *more of the collaborative disciplines.* Brown et al (2015) note that interdisciplinary
268 work is not just for senior people and that many of their collaborators maintained
269 high disciplinary publication rates while being part of a team. On the other hand,
270 it took nearly five years for them to begin publishing their interdisciplinary work in
271 “high impact (i.e. tabloid)” journals².
- 272 • *Nurture constructive dialogue.* To which you might respond “Duh”, but it is true.
273 Via trial and error, Brown et al (2015) found that it was important to create an en-
274 vironment and behavioral rules in which researchers across all disciplines – some of
275 which are vastly different in their particular norms, technical language, and means of
276 communication – could communicate with each other. This included commitments
277 to use plain English (it is wise for you to buy, read, and re-read Strunk and White

²This is my experience with State Dependent Life History Theory implemented by Stochastic Dynamic Programming – Colin Clark and I, and Alasdair Houston and John McNamara had published many papers in specialized journals before our paper in *Nature* (Houston et al 1988).

278 1979, Dreyer 2019), to respect the norms and standards of other disciplines, and to
279 spend time reflecting on what is working and what is not working in the collabora-
280 tions across disciplines. All of these things take practice, and almost nobody who
281 is new to interdisciplinary work will be good at it from the outset. For example,
282 new collaborators tend to try to dominate discussion and assert the primacy of their
283 discipline; they need to be reined in (and hopefully learn quickly how to behave).

- 284 • *Build bridges between research, policy and practice.* Brown et al (2015) empha-
285 size that establishing connections between researchers, policy-makers, and industry
286 stakeholders is key to developing interdisciplinary collaborations. Contrary to what
287 most academics think, industry rarely thinks in disciplinary silos. Indeed, unlike
288 academia where the individual CV is foremost, in the non-academic setting, the
289 success of teams is crucial. Engaging policy-makers and industrial partners in the
290 design of research program from the outset allows policy and industry partners to
291 provide feedback on the scientific approach and presentation of results.

292 Professional Preparation

293 One can begin a career in Pasteur’s Quadrant at any age – Thomas (2000), Dill (2017),
294 and Ommer (2018) provide interesting and contrasting a personal histories of their de-
295 velopment for work in Pasteur’s Quadrant³. Dill (2017) and concludes that functional
296 approaches, such as those used in behavioral ecology (and State Dependent Life History
297 Theory implemented by Stochastic Dynamic Programming) allow predictions from first
298 principles of the range of possible responses to environmental change outside the range of
299 conditions that organisms have experienced in their evolutionary history or ontogeny. He
300 also notes that models have the strength in that they can be generalized across species
301 and ecosystems. Most of us cannot be the 19th century explorer Richard Burton, who
302 was a one-man interdisciplinary team (Pettitt 2015). But we can get ready for work in
303 Pasteur’s Quadrant.

304 In Figure 3, I show the problem-solving plane that I developed in discussion with Phil
305 E. DePoy when he was president of the Center for Naval Analyses and I was a consultant
306 there.⁴ In this plane, the x-axis represents the degree of difficulty in solving a problem,

³The papers of Dill and Ommer appear the *ICES Journal of Marine Science* “Food for Thought: Luminaries Collection” that I mentioned earlier.

⁴I thank Spiro Stefano, who was my second PhD student, for the visual here. It works much better in a talk!

307 which ranges from low to high. The y-axis represents the degree of structure in the
308 problem, which includes such issues as how clearly is the problem formulated, are there
309 ready methods for solving the problem, and is it clear which of those methods to use?
310 This also ranges from low to high. Beginning in the upper left hand quadrant (Figure
311 3a), a problem with a high degree of structure and a low degree of difficulty is simply
312 boring. Similarly, a problem with a low degree of structure and a low degree of difficulty
313 is irrelevant. In the academic world, we tend to give students problems with varying
314 degrees of difficulty but moderate to high structure – the upper right quadrant in the
315 problem solving plane (Figure 3b). But problems in the actual world usually don't have
316 structure, so they occupy the lower right quadrant of the problem-solving plane (Figure
317 3c). Thus, the applied research world – Pasteur's and Edison's Quadrants – spans the
318 first and fourth quadrants of the problem solving plane, with the vast majority of the
319 problems involving moderate to high degrees of difficulty and moderate to low degrees of
320 structure (Figure 3d). How does one get ready to solve those kinds of problems? The
321 issue is no longer how to solve the problem but how to find the problem in the first place
322 (Mangel 1982, which is hard to find, so I have included it as an Appendix).

323 **Undergraduate and Graduate School**

324 Although it would be ideal to start preparing for a career in Pasteur's Quadrant as an
325 undergraduate, most of us do not. Even if you are already a graduate student, post-doc,
326 or faculty member, I encourage you to read the section on undergraduate preparation.

327 **Undergraduate Preparation**

- 328 • Master Your Disciplinary Major Don't pick a major because it has the fewest num-
329 ber of required courses, but rather because you are willing to delve deeply into it.
330 Most likely your major will have some blend of empirical and theoretical/conceptual
331 traditions (physics is the best in this regard). Learn both sides of that even if you
332 are 'not good in the lab' or 'not good at math' (see below).
- 333 • Get Out Into the Field The best starting point for our kind of work is through an
334 observation that leads you to ask the question "Why is nature like this?". Thus, you
335 should get out into the field. If there are formal field courses, take some. If there
336 are not, create your own by getting out into nature, making an observation, and
337 proposing a hypothesis and asking how you would test it. On my web site, you can

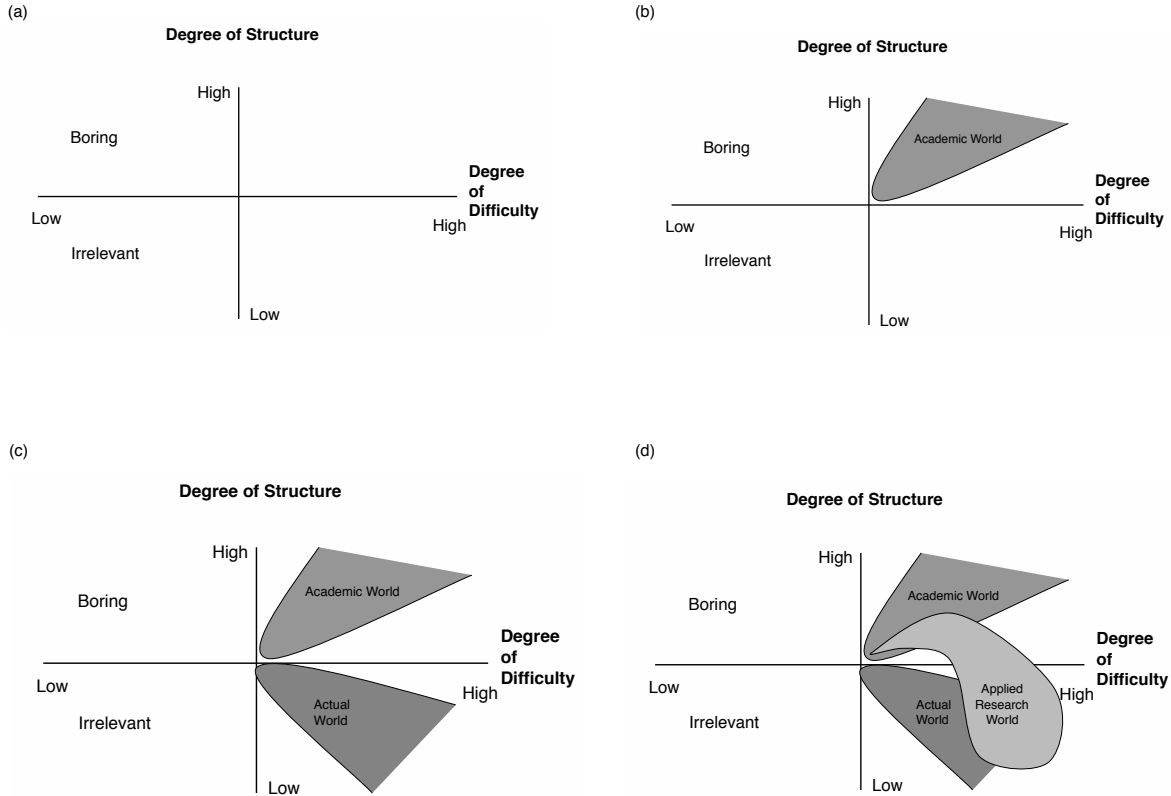


Figure 3: The Problem Solving Plane. a) In this plane, the x-axis represents the degree of difficulty in solving a problem, which ranges from low to high. The y-axis represents the degree of structure in the problem, which includes such issues as how clearly is the problem formulated, are there ready methods for solving the problem, and is it clear which of those methods to use. This also ranges from low to high. problem with a high degree of structure and a low degree of difficulty is simply boring. Similarly, a problem with a low degree of structure and a low degree of difficulty is irrelevant. b) In the academic world, we tend to give students problems with varying degrees of difficulty but moderate to high structure. c) Problems in the actual world usually don't have structure, so they occupy the fourth quadrant of the problem-solving plane. d) The applied research world spans the first and fourth quadrants of the problem solving plane, with the vast majority of the problems involving moderate to high degrees of difficulty and moderate to low degrees of structure. All this begs the question: how does one get ready to solve those kinds of problems?

338 find a field project book that I prepared as a complement to my undergraduate ecol-
339 ogy course that had no laboratory (<https://people.ucsc.edu/~msmangel/field.pdf>).
340 Think it as a set of examples of field projects that you could construct for yourself
341 based on what you have learned in the classroom.

- 342 • Choose Electives to Build Your Interdisciplinary Skills. Among the electives that I
343 believe are most crucial are

- 344 – A course in resource economics, since understanding how costs and benefits are
345 balanced and assessed is crucial. A good example is the new book by Larry
346 Karp on natural resources as capital (Karp 2017).

- 347 – Courses in the social sciences that will help you understand how people make
348 decisions (Kahneman and Tversky 1979, Kahneman 2011, Lewis 2017, Ballard
349 et al 2016, Boyd 2016), how societies function (Converse 2006) and how policy
350 and science advice are passed to governments (Ludwig et al 2001, Hutchings
351 and Stenseth 2016, Milner and Boyd 2017, Murphy and Weiland 2016, Donnelly
352 et al 2018). The best departments for these courses will depend upon the
353 characteristics of the social sciences at your particular college or university.

- 354 – A course in political science, since understanding how government works and
355 policy is made is crucial to what we want to do.

- 356 – A course in art, preferably one that involves active drawing.

357 Why art? There are two reasons. First, artists and scientists have the same problem:
358 to look at a complicated situation in the natural world, decide what aspect of the
359 situation to analyze and communicate, and abstract it for purposes of analysis and
360 communication. Even the best representational art is not, like the best science,
361 ‘realistic’ since the artist, like the scientist, has to decide what to include and what
362 not to include when communicating the idea. Second, the essence of art – as in
363 science – is perception (Edwards 2012) and by enhancing our perception through
364 drawing, we develop a transferrable skill of intuition about other matters.

- 365 • Participate in Research Research is a skill and one develops skills by practice. Colvin
366 (2008), in his book titled *Talent is Overrated*, argued that to be at the top of anything
367 requires 10,000 hours of practice – so get started early and do it consistently. Find
368 ways to participate in research projects as early as you can in your studies. If you

369 can do an undergraduate thesis or capstone project for your major, do it. And try
370 to come up with the idea yourself.

- 371 • Master Your Skills in Modeling and Statistics. Wilson (2013) has very strong opin-
372 ions about whether one has to master quantitative skills. He wrote an essay with the
373 title “Great scientist \neq Good at Math”. Of course, he also published seminal books
374 in mathematical biology (MacArthur and Wilson 1967, Oster and Wilson 1978).
375 Wilson gives two principles: Principle Number 1 is “It is far easier for scientists to
376 acquire needed collaboration from mathematicians and statisticians than it is for
377 mathematicians and statisticians to find scientists able to make use of their equa-
378 tions.” Principle Number 2 is “For every scientist, there exists a discipline for which
379 his or her level of mathematical competence is enough to achieve excellence.”

380 On the other hand, Bangham and Asquith (1991, pg 992) wrote “It is a widespread
381 fallacy that what mathematics contributes to biology is quantification of an oth-
382 erwise innumerate science. But experimental biologists have long been expert at
383 measuring and quantifying. The real contribution of mathematics lies in a precise
384 qualitative framework of reasoning. The rate-limiting step in the advance of biol-
385 ogy is usually experiment, not theory. (One of the very few notable exceptions was
386 the theory of evolution by natural selection.) Experiment, however, is in no sense
387 superior to theory, nor vice versa: both are necessary ingredients of a proper under-
388 standing of nature. An experiment done with no theoretical framework to analyze
389 or interpret the results (let alone a hypothesis) is meaningless; theory in the absence
390 of experiment remains mere theory.”

391 Ellenberg (2014) makes the case that mastering quantitative skills is the easiest way
392 not to be wrong about lots of things. However, you are still likely to make plenty
393 of mistakes if you are working at a level that challenges your abilities; this is where
394 you want to be.

395 I agree with Wilson, Bangham and Asquith, and Ellenberg! I encourage you to
396 master quantitative skills as much as possible. Here’s the reason: with mathematical
397 skills you internalize one level when they are tools at the next level . Putting aside
398 the insipid reason that calculus should be a gate-keeper for medical school, the
399 reason for requiring calculus of biology students is so that they really master the
400 skills of pre-calculus such as knowing what a function is or how to instantly identify
401 the slope and intercept of the line $y = 7x + 5$. These are things that you have to

402 do constantly in research. Imagine if everytime somebody told you that the growth
403 rate of a population is 5% of its current size, you had to write down $dN/ds = 0.05N$
404 and then laboriously solve it, rather than by saying “That means the population
405 grows exponentially according to $N(s) = N(0)e^{0.05s}$ ” because you know this in your
406 gut.

407 There is no doubt that learning quantitative skills is difficult. In another “Food for
408 Thought Piece”, Moore (2020) writes: “Barrier to personal progress in science can
409 also arise from undeveloped skills. ‘Take math courses until you get a C’ is what
410 G. Evelyn Hutchinson, the eminent ecologist (Hutchinson, 2011), told me and other
411 graduate students at Dartmouth College when he visited our small group in the
412 mid-1980s. I have recalled his wise advice numerous times throughout my career,
413 and I passed it on to my students”⁵.

414 **Graduate Preparation**

415 The recommendations that I have about coursework in undergraduate school apply here as
416 well. It is likely to be more difficult to take a variety of courses outside of one’s discipline
417 in graduate school, but I encourage you to find a course on resource economics and one
418 on policy (at least seriously audit them if you cannot take them). Thus, you will have to
419 take more courses to become an interdisciplinary scientist than you would to just become
420 a disciplinary scientist.

421 Similarly, you need to attend meetings in your disciplinary field but at the same time
422 look for opportunities to extend yourself by finding interdisciplinary meetings. When
423 going to meetings, especially the latter ones, remember the Spanish proverb that we have
424 one mouth and two ears to remind of us that we should listen more than talk.

425 You also need to read more widely than you would if you were doing only a disciplinary
426 degree. Avoid the trap of the impact factor and the high profile journals. By its very
427 nature, work in Pasteur’s Quadrant requires more space for the full development of ideas
428 than does work in some of the other quadrants. Thus, you may not find the right kinds
429 of papers in the high profile/high impact factor journals. And if you do, the important
430 material will often be buried in Appendices that are very hard to decipher (that is, it will
431 be difficult to impossible to reproduce the work described in these appendices).

⁵I have not looked up the Hutchinson citation, it is: Hutchinson, G. E. 2011. in *The Art of Ecology: Writings of G. Evelyn Hutchinson*. D. K. Skelly, D. M. Post, and M. D. Smith (editors). Yale University Press, New Haven, Connecticut, USA. 356 pp.

432 **Working Hours and Work Life Balance**

433 Although this section could sit with the previous one, I give its own heading because of
434 the importance of the topic of work-life balance.

435 Over the last few years, when visiting universities to give talks, I have done mentoring
436 sessions with graduate students and post-docs. One question that consistently comes up,
437 especially from those interested in a career in academia, is whether they need to regularly
438 work 70 hour weeks in order to succeed.

439 There are a number of comments I make. First, it is really hard to *work* a 70 hour
440 week. I have done it on occasion but consider that few – if any of us – can sustain that
441 for very long and continue to be productive.

442 Second, there is a threshold-asymptotic relationship between the level of effort that
443 you put in and the quality of the product (Figure 4). That is, there is a minimal level
444 of effort for there to be any quality in the product. Above this threshold we might
445 imagine two models for quality as a function of effort. In the first model, there is a linear
446 relationship between effort and quality until a certain level of effort is reached after which
447 quality remains constant. In the second model, quality is an asymptotic function of effort
448 – you may never reach maximum quality and part of the challenge here is to decide at
449 which point to consider sufficient effort to be put into a product. I give three examples
450 in Figure 4– you could stop at about 60%, 85% (by increasing the effort by fifty-percent
451 over that giving 62%) or 92% (by doubling the effort of that giving 62%) of maximum
452 quality. Deciding where to stop depends on what happens next with the product. For
453 example, for a first draft of a paper with collaborators you might want to stop at 70% of
454 maximum quality (thus getting the draft to them sooner rather than later, and allowing
455 real opportunity for them to improve the paper), with a thesis chapter that you expect
456 serious comments from your committee perhaps 85% of maximum is better, and with a
457 paper that you are about to submit, 90-95% is a good stopping point. Now, if you want
458 to guarantee that you both overwork and rarely get things completed, reason like this: “I
459 am at 92% of maximum possible quality, but if I just put in the same amount of effort as
460 I have already, then I can claim the next 5% of quality.”

461 One of the tricks to both not working 70 hour weeks and getting things out is learning
462 which tasks just need to be done, which need to be done well enough, and which need
463 to be done really well (but not to 100% of maximum quality– which may take you into
464 the asymptotic time sink). For example, a letter to the campus registrar about the poor
465 quality whiteboard does not need to be the best possible letter of complaint, rather it just

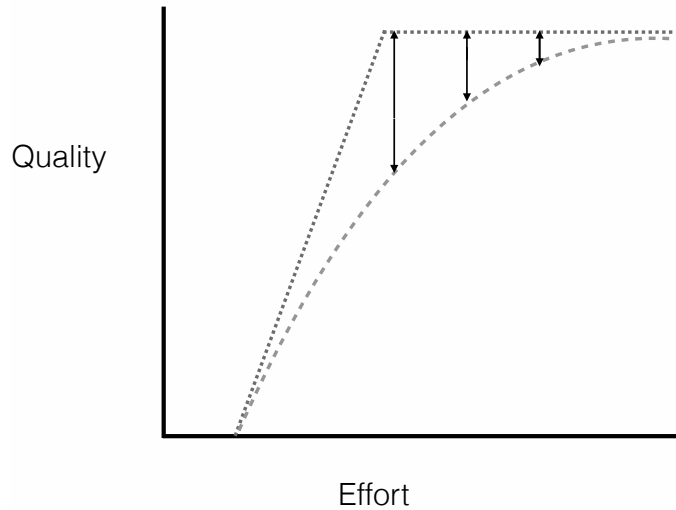


Figure 4: A theory of effort and quality of research and writing. Without some minimum level of effort, there is no quality. After that threshold is passed, quality increases with additional effort. But quality always has a limit (whether that is approached linearly or asymptotically is depends upon the situation, but my opinion is that it is almost always asymptotic). One aspect of having a successful work-life balance is to learn to be satisfied with quality that is close to but below the asymptote.

466 needs to go out. A grant proposal needs to be done really well.

467 Third, as your career progresses, the quality per unit of effort will increase. Now since
 468 quality per effort is a ratio, it can increase in one of two ways. First, it could be that
 469 quality remains the same, but the effort needed to produce that quality declines with
 470 experience. When I was a graduate student, the western novel writer Louis L'Amour
 471 (check spelling) was interviewed on the news show *Sixty Minutes* and he said that he
 472 could write a novel sitting in the middle of the LA freeway – demonstrating his ability
 473 to concentrate (and his fearlessness) and the kind of formula that he had for writing. I'd
 474 say there that quality remained constant but effort declined.

475 The other way that quality per unit effort increases is that the struggle is always
 476 present, but the quality for the same effort increases. In 1991, John Updike published
 477 his collected non-fiction from the 1980s (Updike, 1991). In the preface to this collection,
 478 Updike noted how hard it was for him to write some of these pieces. When John Updike
 479 finds it a struggle to write, we should all take solace and just keep working.

480 Finally, work-life balance is important if you are to be both a successful scientist and
 481 a content person. A 70 hour work week does not allow for any kind of balance. I will
 482 leave you with a homework problem: who impresses you more – the person who works
 483 10 hours a day seven days a week and succeeds or the person who works 6 hours a day 5

484 days a week and succeeds? Which kind of person do you want to be?

485 **Postdoctoral Work**

486 Hans Krebs considered the postdoctoral period to generally be the most important and
487 formative for one's career. He wrote "To attain distinction in scientific research you must
488 1) get a post-doctoral Fellowship, which will give you the time and opportunity to test
489 yourself; 2) attach yourself to a centre of excellence; 3 work hard and make the fullest use
490 of the time and facilities that the Fellowship affords; and from time to time, search your
491 heart critically, with the help of objective critics, to find out whether you really possess
492 the right mixture of qualities – urge, commitment, imagination, humility– which are the
493 roots of creativity in science." (Krebs 1981, pg 179). In a festschrift honoring the John
494 Bachall, William Press (whose article was is about using Bayesian methods with data
495 on the Hubble constant) wrote that Bachalls's advice to young scientists was to choose
496 important problems, get involved with real data, and bring the most powerful analytic
497 tools to bear on the problem at hand (Press 1997, pg 59).

498 So how do we find the important problem? Is it the one that is trendy right now?
499 Maybe, but perhaps not.

500

Richard Feynman's Advice to Koichi Mano on Picking Problems in Science

After Richard Feynman received the Nobel Prize in physics, he received a congratulatory note from Koichi Mano, who had been at Cal Tech when Feynman was doing the work for which he won the prize but who was back in Japan. In his note to Feynman, Mano wrote that he was a nameless man working on unimportant problems. Feynman's poignant response is this:

"The worthwhile problems are the ones you can really solve or help solve, the ones you can really contribute something to. A problem is grand in science if it lies before us unsolved and we see some way for us to make a little headway into it. I would advise you to take even simpler, or as you say, humbler, problems until you find some you can really solve easily, no matter how trivial. You will get the pleasure of success, and of helping your fellow man, even if it is only to answer a question in the mind of a colleague less able than you. You must not take away from yourself these pleasures because you have some erroneous idea of what is worthwhile.

"You met me at the peak of my career when I seemed to you to be concerned with problems close to the gods. But at the same time, I had another Ph.D. student (Albert Hibbs) whose thesis was on how it is that the winds build up waves blowing over water in the sea. I accepted him as a student because he came to me with the problem he wanted to solve.

"...No problem is too small or too trivial if we can really do something about it. You say you are a nameless man. You are not to your wife and to your child. You will not long remain so to your immediate colleagues if you can answer their simple questions when they come into your office. You are not nameless to me. Do not remain nameless to yourself – it is too sad a way to be. Know your place in the world and evaluate yourself fairly, not in terms of ideals of your own youth, nor in terms of what you erroneously imagine your teacher's ideals are." (Feynman 2005, pg 198-199).

502 When looking for a post-doctoral position, seek one in which you will be allowed to
503 develop your own ideas, rather than just work on those of your supervisor. For example,
504 ask if you can have one day a week for your own scientific development that will give you
505 momentum going into your next position.

506 Finally, use the post-doc to continue to develop your skill set (get that course in
507 resource economics now if you have not done so and always learn more statistics and
508 modeling); for example audit one course a semester/quarter.

509 **Find Mentors at Every Point in Your Career**

510 We are not in this alone (Obama 2018). As an undergraduate student you will likely
511 be assigned an advisor. Treat your advisor as a mentor, ask questions of them and seek
512 advice. If your advisor seems uninterested in you, then seek a different advisor. At every
513 level, if you mesh well with a particular instructor, seek them out as a mentor as well. My
514 very first scientific mentor was William Hay, who taught an upper division oceanography
515 course that I took (you can read more about Bill in the *Homage to My Mentors*, to be
516 found on my web site.

517 As a graduate student and post-doctoral scholar you have a supervisor. If the world
518 were perfect, every graduate or post-doctoral supervisor would be a terrific mentor. Sadly,
519 the world is not perfect. Thus, it is on your shoulders to once again find mentors, by
520 meeting faculty other than your supervisor, by asking to attend group meetings of other
521 faculty, and by simply observing how people behave. When you are starting out as a
522 graduate student, more senior graduate students and post-docs can also be your mentor.

523 This process continues once you start a permanent job. Whether in academic or non-
524 academic work, discover the people whom you admire and treat them as your mentor. Of
525 course, as you get older, it may happen that younger people can serve as mentors too.
526 For example, a number of my junior colleagues are my mentors with R and I am happy
527 to have their mentorship.

528 We are truly not alone.

529 **Communicating Science**

It wasn't by accident that the Gettysburg address was so short. The laws of prose writing are as immutable as those of flight, of mathematics, or physics – Ernest Hemingway to Maxwell Perkins

530 If a lecture is not as alluring to the audience as a bride to her groom, you had better not deliver it at all – Simeon Ben Lakish

If I am such a natural athlete, how come I have to practice 11 hours a day – Bill Russell

It usually takes more than three weeks to prepare a good impromptu speech – Mark Twain

531

532 Writing and speaking are also skills. As with all skills, these require some talent and
533 lots of hard work. The ability to communicate science to the non-expert is about as
534 important and pressing as it has ever been. For that reason, I will emphasize speaking
535 more than writing. When writing the first draft of this chapter, I was reading Hope
536 Jahren's *Lab Girl* (Jahren 2016). There is much to recommend about a book that begins
537 "THERE IS NOTHING in the world more perfect than a slide rule" (Jahren, 2016, pg
538 7)⁶. Jahren has been described as a rock-star of science communication, and although we
539 all cannot be like her, we all can improve our ability to communicate science. One of the
540 things that is also apparent very early in Jahren's book is how hard she has worked and
541 works. Writing is an essential part of communicating science (publish or perish happens
542 pretty much everywhere in some form, not just in academia), pretty much almost everyone
543 getting a PhD should understand the importance of written communication. However,
544 verbal communication is equally important but often not given appropriate attention in
545 graduate programs.

⁶Younger readers may not know what a slide rule is or how it is used and Jahren's book really does not do justice to slide rules. I encourage you to put away your phones and calculators, seek out a slide rule, and have some fun with it.

546 In this section, we will explore two points First: No matter how good you are techni-
547 cally, if you cannot tell your story in a compelling manner, it will be lost. Perhaps we need
548 short and pithy advice columns on how to do this rather than whole books; for example
549 in the 22 Nov 2016 issue of *Forbes* magazine, Marshall Shepherd, wrote a column “9 Tips
550 For Communicating Science to People Who Are Not Scientists”. His tips are consistent
551 with much of this chapter, as is the advice of Martinez-Conde and Macknik (2017).

552 The second point comes from former Mangel Lab member Kristen Honey (who is now
553 doing wonderful things concerning Lyme disease – visit her site at Honey2Healing.org).
554 Kristen reminded me that for our group meetings the tradition was everyone – even brand
555 new students – talked about their research (ideas, results, problems) and that my mantra
556 to the group was “Practice, practice, practice”.

557 Kristen went into more detail and wrote to me “What I remember you sharing with
558 us was: to say ‘Yes’ to every speaking opportunity that came along – whether teaching a
559 class/seminar, giving a course lecture, speaking at a conference, a departmental brownbag
560 lunch/seminar, a lab talk, etc. Apply for any and all conferences and opportunities to
561 publicly share your research while in graduate school. Practice speaking in public in front
562 of audiences big or small. Public speaking is a teachable skill, which can be honed and
563 developed as one’s experiences and time practicing add up over the years, and, eventually,
564 decades... to become an excellent speaker in your own style/voice. Practice. Practice.
565 Practice.”

566 I do not recall saying this, but also do not doubt that I did. Kristen also wrote to me
567 that the advice about speaking is “analogous advice to how you used to say ‘Math is not a
568 spectator sport. Think of it like going to the gym to build muscle strength with sustained
569 commitment and applied effort. Keep doing math – your own math – not just watching
570 others (nobody ever got strong by watching other people work out in the gym) – and
571 build your math muscle.’ The common theme about both math and speaking: Practice.
572 Practice. Practice. Both are teachable skills that improve with time and commitment.”

573 **Writing**

574 Skills improve more rapidly with effective coaching and consistent training. For writing,
575 this means both sharing what you have written and trying to learn about how to be a
576 better writer. To do the latter, it is helpful to have a collection of books on writing. Here
577 are some of mine:

- 578 • Alley (1987) *The Craft of Scientific Writing*
- 579 • Atwood (2002) *Negotiating With the Dead. A Writer on Writing*
- 580 • Day (1998) *How To Write and Publish A Scientific Paper. 5th Edition*
- 581 • Dreyer (2019) *Dreyer's English*
- 582 • Elbow (1998) *Writing With Power*
- 583 • Evans (2017) *Do I Make Myself Clear? Why Writing Well Matters*
- 584 • Freed (2005) *Reading, Writing, and Leaving Home. Life on the Page*
- 585 • Friedland and Folt (2000) *Writing Successful Science Proposals*
- 586 • George (2004) *Write Away*
- 587 • Greene (2013) *Writing Science in Plain English*
- 588 • Higham (1998) *Handbook of Writing for the Mathematical Sciences*
- 589 • Pinker (2014) *The Sense of Style. The Thinking Person's Guide to Writing in the*
590 *21st Century*
- 591 • Prose (2006) *Reading Like a Writer*
- 592 • Rabiner and Fortunato (2002) *Thinking Like Your Editor*
- 593 • Schimel, J. *Writing Science. How to Write Papers that Get cited and Proposals that*
594 *Get Funded*
- 595 • Silva, P.J. *How To Write a Lot*
- 596 • Strunk and White (1979) *The Elements of Style*
- 597 • Truss (2003) *Eats, Shoots and Leaves*

598 You should not rush out to buy these; I have collected them over a long period of time.
599 One of these authors writes fiction; you may ask what can you – a scientist – learn from
600 them? Plenty. For example, Elizabeth George writes fiction about a British detective
601 named Thomas Lynley. Her book about the writing life is roughly 50% about how to
602 write fiction and 50% about the habits of writing, which is why it is on this list. She also

603 makes the point that one needs to ask if he or she wants to be an author – somebody
604 who gets up every day and goes to the desk to write – or an Author – someone who
605 goes on TV shows, is photographed for profiles magazines, and does all the other things
606 that prevent one from writing. Similarly, you might ask if you want to be a scientist –
607 somebody who does science every day – or a Scientist – somebody who traipses around
608 the world speaking about the work that other people do for him or her.

609 Gopen and Swan (1990), Plaxco (2010), Heard (2014), Yossa (2014), Endler (2015),
610 Grossman (2017), and Mensh and Kording (2017) are papers about writing papers.

611 Among these books and papers, which are all worth reading and re-reading, I take
612 special note of Strunk and White (1979), which you definitely should buy, read and re-
613 read regularly. Pinker (2014) and Dreyer (2019) each offer an update of and homage to
614 Strunk and White.

615 Tim Smith told me that when he began to write his wonderful book on the history of
616 fishery science (Smith 1994), a friend of his who is a writer told him that the author of
617 a book is in a struggle with the reader, because the reader can always decide to put the
618 book down and never open it again. The challenge of the writer is to keep them reading.
619 Similarly, even if it is only for five minutes, the challenge for the speaker is to keep the
620 audience engaged, rather than wandering off in their daydreams.

621 Think that you are building a relationship with your reader. It is conversation, but
622 does not happen in real time.

623 The most important message of these books and articles on writing is this: Get yourself
624 into the writing habit. One can read books about writing forever and not do it. Here's a
625 simple but important calculation. Imagine that you write for one hour a day, five days a
626 week, forty weeks a year so that you give yourself nearly an entire quarter off from writing.
627 If you can write one page in that hour you will write $1 \cdot 5 \cdot 40 = 200$ pages a year. Certainly
628 then, especially early in your career, you will then run out of things to write about long
629 before you run out of time for writing. So where is the difficulty? Finding that 1 hour.
630 Different people will resolve it in different ways. I like to write first thing in the morning,
631 when my mind is freshest and least encumbered. Other people like to write at the end of
632 the day, when they have had time to synthesize. Some people prefer to do their 5 hours
633 in one block or two 2.5 hour blocks, although my opinion is that energy fizzles with such
634 long effort. Find what is right for you and then stick to it – this means writing at least
635 every work day (a friend of mine from undergraduate school who is science fiction author
636 writes 6 or 7 days a week). In his Food for Thought piece, Cury (2018) recommends

637 writing every day; he does so for two hours with no distractions. Try it out.

638 The sticking to it is very important. Many Hemingway wanna-bees forget that no
639 matter what the great writer did the night before, the next morning he was at his table
640 in the cafe, pencils and yellow pad in hand.

641 **Speaking: The Role of Deliberate Practice**

642 There are also literally hundreds of full books written about communicating science. Two
643 nice recent ones are Dean (2009) and Walker (2010)). Dean is a science journalist and on
644 the faculty at Harvard. This is a wide ranging book for scientists about communicating
645 with the public and covers the landscape in which such interactions may occur. The second
646 chapter is “Know Your Audience”. This is a very handy book once one has mastered the
647 art of the short talk, a good reference when a journalist is calling, or when one acts as an
648 expert witness.

649 Walker runs a series of companies on media training. This book goes into much more
650 detail about technique, but pretty much begins at the time of the presentation, so that
651 there is little about deliberate practice.

652 We will begin with strategy and tactics for making a short a success. We will work
653 towards deliberate practice for communicating. The rubric some of us learned in high
654 school speech class –“Tell them what you are going to tell them, tell them, and tell
655 them what you told them” – is a good start, but insufficient. We should seek to identify
656 the laws that Hemingway wrote about to Maxwell Perkins, and modify those laws for
657 communicating with non-experts?

658 Since speaking is a skill, each one of us is born with a different level of ability to give
659 talks, but without deliberate practice one does not hone the skill; talent is once more
660 overrated (Colvin 2008). Colvin is a senior editor for *Fortune* magazine and the subtitle
661 of his book is “What Really Separates World-Class Performers from Everybody Else” and
662 the answer is deliberate practice; he explains how to conduct deliberate practice in great
663 detail. One can spend all of one’s time reading about how to prepare short talks without
664 actually preparing any; this will not do you much good. Regardless of whether you are
665 having a 5 minute talk about what you do at work with your grandmother or giving a 50
666 minute talk in a job interview, you must tell a story. So let’s get going.

667 **The Short Talk: Elevator Speeches and 5 minute Talks**

668 By a short talk, I mean one of 5-10 minutes given by one person to a few to many other
669 people. Almost *everyone*, not just scientists, does this in the course of their work. Short
670 talks include contractors explaining to home owners the details of a potential remodel;
671 ministers, rabbis, and imams speaking to their congregations; entrepreneurs speaking to
672 venture capitalists either formally or in an “elevator speech” (if you – the entrepreneur
673 – have not captured the venture capitalist in three to five minutes, go back and work on
674 your pitch some more). Even much of teaching consists of short talks – in which questions
675 are answered for small groups of students who are attentive at that moment.

676 Intentionally being redundant: Giving such talks is a skill, which means that no matter
677 how much native talent one has, the quality of the talk can be improved through deliberate
678 practice. In general, all talks involve asymmetric information, since the the speaker
679 knows much more about the subject than the audience; the great challenge is how to
680 communicate the key ideas most effectively.

681 You should aim to communicate: **W**hat you are working on, **W**hy you are doing it,
682 **W**hich tools you are using, and **W**isdom you have gained. Think of these as **W**⁴.

683 *Know Your Subject:* One cannot give a good and meaningful short talk without know-
684 ing the subject matter. This is more than accumulating a collection of facts, but collecting
685 the facts and organizing them in a way that is consistent with the goal of the talk. As
686 you read, it is key to allow sufficient lead time for subconscious work along with conscious
687 work.

688 *Know Your Audience:* In most cases we have an idea of who the audience is and need
689 to think carefully about the limitations of their knowledge. Imagine a contractor talking
690 to another contractor rather than to a client who who has never built anything. Imagine
691 explaining “where babies come from” to a 5 yr old, 10 yr old or 15 yr old.

692 *Learn More and Use Less – Selecting Your Knowledge to Match Your Audience:*
693 The real secret to the short talk is figuring out how to reach your audience without
694 overwhelming them. In almost every case, this means using less information than you
695 have at hand. Thus, we need to winnow information to create the optimal match between
696 the needs of the audience and the knowledge of the speaker. The focus is finding a nugget
697 of information that they can hold onto after the talk ends.

698 **Technique Matters**

699 Even a good match between the audience and your knowledge base is not sufficient,
700 technique is important. Important aspects of technique are:

- 701 • Speak clearly and slowly: We all know how annoying it is to try to follow a speaker
702 who speaks too quickly, mumbles, is not articulate, or does not project. Don't be one
703 of those people. If this means that you need to use a microphone where somebody
704 else might not, use the microphone. Experiment with its placement so that you get
705 to know it before you have to start to use it. We all tend to speak too quickly when
706 we are excited about a topic or nervous in front of an audience, so be sure to slow
707 down. In fact, if you use written notes, at the top of them it is a good idea to write
708 "SLOW DOWN!!"
- 709 • Move while talking in a way that adds rather than distracts: Speaking is somewhat
710 like acting, with the important difference that an actor does the same thing for a
711 different audience every night, but we (especially when teaching) do different things
712 for the same audience. Actors learn is how to move their bodies in a way that the
713 visual effect of movement enhances the aural input. Think about this as you imagine
714 your body motions during a talk.
- 715 • Address and engage the audience: The best way to do this is to look at the people in
716 the audience. Scan the audience, to assess if people are following you, and sometimes
717 directly look at an individual (or two or three) for a more detailed sense of how you
718 are being followed.
- 719 • With great Power(point) comes great responsibility: One might say that "Power-
720 point does not bore people, people bore people" but that is only partially true.
721 When using Powerpoint think about the information content per slide, number of
722 slides per minute, and how are you going to use the slides. Are you going to face the
723 slides and read them (do not do that), point out salient features, or say in different
724 words the information that is on the slides (thus providing both aural and visual
725 sources for the same information).
- 726 • Chalk or White Boards: It is unlikely (but not impossible) that you will face using
727 a chalk board (some are still around), but you are likely to use white boards espe-
728 cially when speaking extemporaneously. Writing on a board is another skill – which

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means you need to practice how big the letters are, figuring out the best density of information per board, and how to simultaneously write on the board and how to talk to the audience. For the latter, turn to face the audience frequently – saying, whenever possible, what you have just written but in a different way, once again giving visual and aural input for the same information

- Notes: Most of us will not talk with tele-prompters, but we may use notes that range from from notecards with key words to a few pages with key ideas to notes on your PowerPoint or Keynote file to a fully written talk that you read. What you use is a matter of personal choice – but you have to experiment (i.e., practice) to figure out which works best for you and how not to bury yourself in the notes.
- How You Dress Is Important: The way that you dress gives the audience a sense of your respect for them. Thus, aim to mirror the dress standard of the audience (even if it means asking in advance). It is better to overdress than underdress. Nobody is going to fault you for wearing a tie to a job interview, even if none of the audience (except possibly the Dean) do, but if you show up in a tee shirt and torn jeans (thinking that this will show what a great scientist you are) you have a battle from the start – people will be asking about why you dressed that way, rather than listening to what you are saying.
- Don't be Arrogant. It is easy for highly trained people to become arrogant and play 'gotcha' with an audience, to show off to the audience. Here's an example, which I used both in *The Ecological Detective* (Hilborn and Mangel 1997) and *The Theoretical Biologist's Toolbox* (Mangel 2006) because it tells the story so perfectly. It comes from Hogarth (1980).

Consider these two series:

Series A: 45, 32, 12, 23, 26, 27, 39

Series B: 1040, 1027, 1007, 1018, 1021 , 1022, 1034

When asked which sequence is more variable, most people will say that sequence A is more variable than sequence B. However, sequence B is sequence A plus 995, so that the variance of these two sequences is exactly the same. Hogarth writes: "Which series exhibits more variability? Most people answer series A. However, the statistical measure of variance - which indicates the amount of irregular variations

761 from the mean of a series of numbers - is the same for both series. Series B is simply
762 series A plus a constant. However, intuitive judgments of variability are usually
763 influenced by the size or context of the series or objects. That is subjectively relative
764 variability is more salient than variability per se” (Hogarth 1980, pg 44).

765 However, variance or standard deviation by itself, lacking units, is nearly meaning-
766 less. Is a standard deviation of 5 large? Who knows? Compared to what? Without
767 units we cannot assess the meaningfulness of the standard deviation. The coefficient
768 of variation, standard deviation divided by the mean, has no dimensions, so if we
769 have a coefficient of variation that is 5, we then can conclude it is large, since this
770 is telling us that the standard deviation is 5 times the mean. The coefficient of
771 variation of sequence B is much smaller than that of A. You should (i) verify that
772 this is true by computation, and (ii) understand the reason for this being true.

773 In *The Theoretical Biologist’s Toolbox*, I added a series and made it into an exercise
774 that is both easy and fun (Mangel 2006, pg 88). I repeat it here. We now have

775 Series A 45, 32, 12, 23, 26, 27, 39

776 Series B: 1401, 1388, 1368, 1379, 1382, 1383, 1395

777 Series C: 225, 160, 50, 115, 130, 135, 195

778
779 Ask at least two of your friends to, by inspection, identify the most variable and
780 least variable series. Also ask them why they gave the answer that they did. Now
781 compute the mean, variance, and coefficient of variation of each series. How do the
782 results of these calculations shed light on the responses?

- 783 • Spend Time Thinking About Different Ways of Presenting the Same Information. Those
784 of you who have taught know that much of being a good teacher is anticipating the
785 different ways that students may not understand the material, and coming up with
786 ideas that will help overcome the barriers. That is why teaching takes so much sub-
787 conscious time (which is also the time spent on research – one of the reasons that
788 we are often less productive in the midst of teaching).

789 This same problem arises when we give talks, and it is worthwhile for you to spend
790 time in advance thinking about different ways of presenting the same information.
791 I will give two examples.

792 From 2013-2016, I served on the Scientific Review Board of the International Pacific

793 Halibut Commission (you can find out about us here <http://www.iphc.info/srb>).
794 The SRB's role is to meet with the Commission scientists to provide feedback about
795 both the science that they do and communication of the science to the Commission-
796 ers.

797 One thing we discovered is that people understand a phrase such as “6 times out of
798 100” more readily and fully than simply saying 6% of the time. I am not certain
799 why this is so, but it could be that when we say 6 times out of 100, it is easy
800 to envision 100 coin flips, or bottles into which we are dropping marbles (or your
801 favorite visualization) than the more abstract 6%.

802 We also found that concepts about probability are more effectively communicated
803 as cumulative distributions rather than probability densities. Imagine trying to
804 communicate the probability density of spawning biomass. First, we have to explain
805 the notion of a continuous random variable X and that the probability density $f(b)$
806 is interpreted to mean that the chance that biomass B falls between b and $b + db$ is
807 approximately (remember those pesky $o(\Delta b)$ terms when defining a density (Mangel
808 2006)) $f(b)\Delta b$. Thus, we would illustrate that the chance of the biomass being
809 about the specific value b_1 and a bit more than that Δb is $f(b_1)\Delta b$. To describe
810 a confidence interval, we need to implicitly discuss the notion of integrating the
811 probability density.

812 If instead that we use the cumulative distribution function, we explain that this
813 function $F(b)$ represents the chance that B is less than b – a much simpler concept
814 to convey, with a plot that only rises in the b - $F(b)$ plane. Furthermore, we can easily
815 illustrate the ideas of the 5% and 95% confidence interval by drawing horizontal lines
816 from 0.05 and 0.95 and seeing where they intersect the curve. The point here is not
817 that you should use cumulative distributions rather than probability densities or
818 that you should report 6 times out of 100 rather than 6%. In summary, you should
819 continuously think about better ways of communicating quantitative concepts and
820 think of communicating as an experiment – try things out, assess what works and
821 does not work, and modify accordingly.

822 **Using Practice to Improve: The NW to SE Assessment of Skills**

823 We now come to the process of deliberate practice, ranging from actual practice talks to
824 thinking hard about the talk one is about to give, and its role in making your short talks

825 excellent.

826 To the old saw “Practice makes perfect” one can respond “Practice makes permanent”
827 – that is simply practicing may reinforce your bad habits. The way to get around this is
828 through feedback.

829 Here’s a feedback sheet that I used in my lab group meetings when students, post-
830 docs, or I practiced talks, and in our team taught course for graduate students in their
831 first quarter. The compass headings NW and SE indicate Needs Work and Skilled and
832 Effective respectively.

833 Name of Speaker:

834 *General skills*

835 Voice:

836 Clarity NW == == == == == == == == SE

837 Pace NW == == == == == == == == SE

838 Volume NW == == == == == == == == SE

839

840 Carriage:

841 Confidence NW == == == == == == == == SE

842 Mannerisms NW == == == == == == == == SE

843

844 Engagement:

845 Eye contact NW == == == == == == == == SE

846 Use of repetitive phrases NW == == == == == == == == SE

847 (eg. Um, like, so, you know)

848 Kept within time limits NW == == == == == == == == SE

849

850 Scientific skills:

851 Motivated the question: NW == == == == == == == == SE

852 Math Presentation: Level of detail NW == == == == == == == == SE

853 Math Presentation: Clarity NW == == == == == == == == SE

854

855 Information Flow:

856 Sequencing NW == == == == == == == == SE

857 Major Topic Emphasis NW == == == == == == == == SE

858

859 Slides:
860 Number NW == == == == == == == == SE
861 Detail NW == == == == == == == == SE
862 Interest NW == == == == == == == == SE

863
864 Content:
865 Introduction: Clear and concise NW == == == == == == == == SE
866 Body: Expanded introduction NW == == == == == == == == SE
867 Conclusion: Repetition of main themes NW == == == == == == == == SE

868
869 *General comments:*

870 An important point about using a sheet like this: it is best done in a situation in
871 which over the course of the afternoon, weekly group meeting, or formal course, everyone
872 speaks. This helps to temper comments – when writing comments on others, one knows
873 that his or her turn is coming and that tends to make everyone better behaved to know
874 that they too will talk at some point.

875 **The 12-15 Minute Talk**

876 Although there are instances of 5 minute talks in professional settings, most of the short
877 talks that we give as scientists occur in society meetings with 12 or 15 minutes allocated
878 for your talk, questions, and transitions (so think of these as 9 to 12 minute talks).

879 One can view giving a talk as a research problem. Construct hypotheses about how
880 to effectively present the information and then try them out. Regardless of how it goes,
881 analyze the result. If you were successful, you want to know why and if the talk bombed
882 you also want to know why. Trust me – we all give talks that bomb.

883 Here there are a number of other things to consider

- 884 • Detail versus flow and clarity: The essence of dynamic programming is thinking
885 backwards in time and you can do much the same with planning your talks. That is,
886 ask yourself “what do I want people to remember when I finish the talk?” and then
887 ask how you can get there. Doing so will allow you to construct the flow of your talk
888 from the targeted end point to the very beginning. In practice, you might even begin
889 by writing your conclusions slide. When doing this, every time you add a detail you
890 can check whether or not it will contribute to the flow to the conclusions and the

891 clarity of what you are doing. If it does not, cut that detail. Once you have reached
892 the introductory slide, you can now iterate forward – working on clarity and focus –
893 to get the detail of flow and clarity as best as possible. This same kind of backward
894 and forward iteration is used to solve dynamic state variable games (Alonzo and
895 Mangel 2001, Alonzo et al 2003).

- 896 • Short Term Memory and the Aural/Visual Focus: Psychologists tell us that peo-
897 ple can keep no more than 7 items in short term memory (Miller 1956), although
898 Marshall Shepherd, whom I mentioned above, says that it is closer to 3 items. The
899 point is that your audience will only remember a small amount of what you are
900 saying (and if you are in a session with 4-6 other talks, only a small amount of what
901 was said in total). Keep that in mind as you trim. Also, as we have discussed above,
902 keep working on using your slides so that the aural and visual reinforce each other –
903 use your spoken words to enhance the visual of the slide and vice versa. Never read
904 the slide to the audience.
- 905 • Dealing with Technical Questions: We are all absorbed, and often motivated, by the
906 details of our work. And once in a while, you will get a similarly interested person
907 asking a question following a talk. Resist the temptation to get into the details when
908 you are standing up and everyone is watching – rather suggest that you meet after
909 the talk (e.g. at the end of the session for a coffee) to discuss those details.
- 910 • Saying “I don’t know”: At other times you will be asked a hard question to which
911 you do not know the answer. When that happens, do not try to bluster your way
912 through it – most of the audience will see what you are doing and think less of you
913 rather than more for trying this. Sometimes, in fact, a question to which you do not
914 know the answer may lead to a publication. This happened to me at a workshop on
915 extinction risk for west coast salmonids in 1996, when I was asked a question from
916 somebody way back in the audience and did not know the answer. I replied “I do
917 not know – we should write a paper on that”, which we did (Cooper and Mangel,
918 1999).
- 919 • Making Equations Interesting: Even in a short talk, there are times when you will
920 need to show equations. The path bifurcates very quickly. If the equations are too
921 difficult to explain in a small amount of time (which is all that you have in a short
922 talk), then simply show them and provide a reference to the equation. The idea here

(a) *Evaluating Fitness and the Method of Stochastic Dynamic Programming*

$F(x,t)$ = maximum expected accumulated number of potential grandchildren from t to T given that $X(t)=x$

$$F(x,t) = (1 - \lambda)e^{-m}F(x,t+1) + \lambda \max_c \{f(c) + e^{-mh(c)}F(x-c,t+h(c))\}$$

(b) $F(x,t) = (1 - \lambda)e^{-m}F(x,t+1) + \lambda \max_c \{f(c) + e^{-mh(c)}F(x-c,t+h(c))\}$

(c) $F(x,t) = (1 - \lambda)e^{-m}F(x,t+1) + \lambda \max_c \{f(c) + e^{-mh(c)}F(x-c,t+h(c))\}$

Figure 5: One approach to presenting equations in a short talk, taken from a talk that I give on the evolutionary ecology of offspring size and investment. a) I show the dynamic programming equation for the fitness of a pro-ovigenic parasitoid. b) I then circle [in a bright color] one term on the right hand side of the equation – corresponding to the parasitoid not finding a host in period t and explain each of the terms in the product. c) I then circle the second term on the right hand side of the equation – corresponding to the parasitoid encountering a host in period t and then ovipositing a clutch of size c – and explain each of these terms. The visual and aural input thus complement each other.

923 is that those who already know the equation will immediately grab it and those who
 924 do not will have a source to follow up on the ideas.
 925 If the equations are not too difficult to explain – by which I mean that you can
 926 derive them while speaking – then you need to think carefully about how to present
 927 them so that the listeners do not get lost in the weeds. In this case, I have found
 928 two approaches to be successful. One can show the equation and then step through
 929 it, circling various pieces of it (visual) while describing the meaning of those pieces
 930 (aural); I give an example of this approach in Figure 5.

Alternatively, even in a short talk, we may choose to derive an equation, particularly if you believe that the audience has the mathematical chops to follow it. For example, if we were deriving the Holling disk equation, we would first introduce the variables T , λ , and h as the time interval during which foraging occurs, the rate of encounter with the prey items, and the handling time of the prey items, with the goal of discovering how encounter rate and handling time determine the number of prey items encountered between 0 and T . We then introduce S and H as the total search

and handling times, so that the analysis proceeds as follows

$$S + H = T$$

$$H = \lambda Sh$$

$$S + \lambda Sh = T$$

$$S = \frac{T}{1 + \lambda h}$$

$$\text{Number of prey encountered in } (0, T) = \lambda S = \frac{\lambda T}{1 + \lambda h}$$

931 The associated verbal cues would be something like this:

932 “When we sum the search and handling times we get the total time.”

933 “The handling time is the number of prey encountered (λS) times the handling time
934 (h) per prey item.”

935 “We use that in the first equation.”

936 “We now solve for S .”

937 “The number of prey items encountered is then λ times S ”.

938 Depending on the next steps, one might point out that the rate of prey encounters
939 overall is this result divided by T or that the rate of energy accumulation is this
940 result divided by T times the energy obtained per prey item.

941 When making a set of slides with this derivation, my recommendation is that the
942 equations appear one at a time and that only the most recent equation appears in
943 black, with the others in grey (so that they are visible but it is clear which equation
944 dominates focus). This allows the audience to focus on your most current aural cues
945 while having weighted visual cues (black being more important than grey).

946 Remember, of course, that there will be some people for whom mathematics can
947 never be interesting, but only sterile; it is their problem, not yours.

948 **The 50 Minute Talk**

949 We now come to the longest talk that most of us will give (except possibly for teaching
950 a 2 hour block). Pretty much everything that we have discussed until now applies, but
951 there are a few other considerations

- 952 • More time does not release you from considering time constraints: When I started
953 interviewing for jobs towards the completion of my PhD, our friend Chris Friedrichs,

954 a professor of history at the University of British Columbia, gave me advice that
955 I have continued to pass on. It is: *Do not run over*. If you are told that the 50
956 minute talk includes 7 minutes for questions, then hone your talk to be 43 minutes,
957 not 47 minutes. There is no better way of losing your audience, and possibly the
958 job, especially with an afternoon talk, than going over the time limit.

- 959 • You can get into detail: A 50 minute talk allows you to get into detail, but a few
960 experiences will show you that 50 minutes is way too short a time to allow you to get
961 into all of the details that you want. Thus you must still think carefully about which
962 details are essential to your presentation and which you can simply mention (aural
963 cue) and then show (visual cue) a citation for the relevant work. Knowing into which
964 categories details fall depends on knowing your audience. If you are giving a talk
965 involving DNA and mention the Central Dogma, you will treat this differently with
966 an audience of molecular biologists, applied mathematicians , or general public.

967 **The Far Beyond**

968 Wenow briefly touch on the choice of kind of career (academic versus non-academic being
969 the first point of bifurcation), the size of the research group you find yourself in, and
970 funding.

971 **The Impostor Syndrome, The Importance of Stupidity in Science, and Post- 972 Decision Regret**

973 All of us remember the time when we had started calling ourselves by our scientific
974 discipline, as in “I am an ecologist” or “I am a mathematical biologist”, but thinking that
975 we were really not and somebody was “going to find us out”. This is called the impostor
976 syndrome, a term coined in 1978 by two psychologists (Clance and Imes 1978). Avoiding
977 a technical, psychological definition, it goes something like this: even though you have
978 been successful by the usual metrics, you somehow are convinced that you don’t belong
979 here (wherever here is), that you are some kind of a fake or fraud. You may even find
980 ways of explaining your success as good luck or people not really paying attention.

981 When we first encounter the impostor syndrome, we think that with time, it will pass
982 and we will feel more comfortable as a scientific professional. But it does not; more
983 importantly *it should not*. To advance in science as individuals we must push to work just

984 below the limits of our abilities (this is why I dissected “good at math”) – Peter Medawar
985 called this the art of the soluble (Medawar 1982). The idea is that nobody gets rewarded
986 for trying something and consistently failing, even though it is almost always true that
987 some failing is essential for the road to success.

988 Sometimes, we will jump over that limit and try something that does not work out. So
989 get used to it – it means that you are doing the right thing. Schwartz (2008) emphasizes
990 the same when he points out the importance of stupidity in science. The great physicist
991 Paul Dirac is reputed to have said that there are four stages following a discovery. First,
992 you think it is wrong. Second, you think it is right but everyone knows it. Third, you
993 discover that everyone does not know it and you have made a contribution. Fourth, you
994 berate yourself for all the wrong turns you took and all the mistakes you made on the
995 way to the discovery.

996 In summary, if one works around the limit of one’s abilities, then the impostor syn-
997 drome will never go away, and that’s the way you want it to be.

998 This is as good a place as any to discuss another psychological phenomenon called
999 post-decision regret, which is also called cognitive dissonance theory (Festinger 1962).
1000 Again, with as little technical jargon as possible, this means that whenever you make a de-
1001 cision – no matter how carefully you have thought through it – after you make the decision
1002 you will feel that you have made the wrong choice. Whatever the deeper reasons, it seems
1003 that the proximate cause is clear: in making a decision, a choice, you have closed off one
1004 path in favor of another, and one will always wonder about the other path. Robert Frost
1005 summarized it beautifully in his poem ([https://www.poetryfoundation.org/poems/44272/the-](https://www.poetryfoundation.org/poems/44272/the-road-not-taken)
1006 [road-not-taken](https://www.poetryfoundation.org/poems/44272/the-road-not-taken)).

The Road Not Taken

by Robert Frost

Two roads diverged in a yellow wood,
And sorry I could not travel both
And be one traveler, long I stood
And looked down one as far as I could
To where it bent in the undergrowth;

Then took the other, as just as fair,
And having perhaps the better claim,
Because it was grassy and wanted wear;
Though as for that the passing there
Had worn them really about the same,

And both that morning equally lay
In leaves no step had trodden black.
Oh, I kept the first for another day!
Yet knowing how way leads on to way,
I doubted if I should ever come back.

I shall be telling this with a sigh
Somewhere ages and ages hence:
Two roads diverged in a wood, and I?
I took the one less traveled by,
And that has made all the difference.

1007

1008 **Academic or Non-academic Positions?**

1009 The Science Council (<http://sciencecouncil.org/about-us/>) is a broad-based organization
1010 of professional bodies and scientific societies. One of their web posts
1011 (<http://sciencecouncil.org/about-science/10-types-of-scientist/>) lists 10 kinds of scientists:
1012 1) the business scientist, 2) the communicator scientist, 3) the developer scientist, 4) the
1013 entrepreneur scientist, 5) the explorer scientist, 6) the investigator scientist, 7) the policy

1014 scientist, 8) the regulator scientist, 9) the teacher scientist, and 10) the technician scien-
1015 tist. I list these without going into detailed descriptions, which can be found on the web
1016 site, to remind you that there are many different pathways for one's career.

1017 A career in Pasteur's Quadrant can be consistent with any of the kinds of scientist
1018 identified by the Science Council. One should start a career in Pasteur's Quadrant as
1019 soon as possible. That is, don't say "Once I have achieved - - [fill in the dashes], I will
1020 start doing important applied work". Rather find an important applied problem early in
1021 your career, build you disciplinary skills while seeking fundamental understanding when
1022 solving this problem, and implement the solution in an interdisciplinary manner.

1023 Should one choose an academic career? I believe that there is only one compelling
1024 answer to that question: because you feel that you have to teach. This does not mean
1025 i) you don't mind teaching, ii) like teaching, or iii) that you want to supervise research
1026 students. Rather it means that part of you will feel empty if you are not teaching formal
1027 courses, i.e. that teaching formal undergraduate courses is part of the essence of who you
1028 are.

1029 Although I knew this about myself at a very early age, I only figured out how to
1030 articulate the point about teaching being part of my essence towards the end of my
1031 career. All of us have been in a classroom when the instructor did not want to be there,
1032 and know how quickly the students sense this. Similarly, many (perhaps not all) of us
1033 have had very demanding instructors who clearly cared about the class; the students
1034 generally accepted the demands with good cheer. I chaired three departments in the
1035 University of California and sometimes had colleagues tell me that they received poor
1036 teaching evaluations because they were so demanding; however they clearly emanated a
1037 sense of not wanting to be in the classroom at al. Being demanding was simply an excuse
1038 that they used.

1039 If you feel that you do not have to teach, or work on a very specific organism or
1040 system, then work outside of academia has the promise to be as exciting and satisfying as
1041 within academia and much less stressful, with higher pay, and a potentially more collegial
1042 environment than within academia. Indeed, since most faculty do not have experience
1043 outside of academia, they think that outside of academia one has "no freedom" and it is a
1044 cutthroat world. But, as one of my post-docs, whose partner worked outside of academia,
1045 noted to me many years ago, in academia the focus is the individual CV, whereas in
1046 non-academic setting the focus is the team. That is, in academia one advances or not
1047 according to the individual CV, whereas in non-academic settings, one advances according

1048 to whether or not the team succeeds.

1049 The freedom to choose a particular research project is surely greater in academia,
1050 especially if one does not seek external funding. But once you have received grant or
1051 contract for a project, you are constrained to work more or less on that topic – at least
1052 for the length of the grant or contract. Thus, the nature of academic freedom is perhaps
1053 over-stated. Of course, if you really want to work on mating strategies in preying mantids,
1054 then a job at a fisheries agency is probably not a good match; on the other hand if what
1055 excites you is mating behavior and its effect on population dynamics, then a job at a
1056 fisheries agency could be a very good fit.

1057 Doing interdisciplinary work still faces more challenges in academic settings than non-
1058 academic ones in large part because the structure of universities silos disciplines (regardless
1059 of the amount of talk about interdisciplinary work). Cronin (2015) noted that his univer-
1060 sity enabled “rewarding transdisciplinary work”, where rewarding is used as an adjective,
1061 but that it was still resistant to rewarding transdisciplinary work, where rewarding is used
1062 as a verb. As he noted, this remains problematic for those doing interdisciplinary work.

1063 **Leaving and Returning to Academia**

1064 There is great virtue in obtaining experience outside of academia. Thus even if you are
1065 certain that you do need to teach, consider spending time outside of academia. This raises
1066 the question of how one returns to academia. You will almost surely take a cut in salary
1067 to leave a permanent non-academic position to an academic one (my salary went down by
1068 nearly 50% when I left the Operations Evaluation Group of the Center for Naval Analyses
1069 Mangel (1982a) for UC Davis in 1980).

1070 As of this writing, the gold currency in academia is still peer-reviewed publications,
1071 so if you think that you want to return to academia keep publishing even in your non-
1072 academic position. Everyone will understand if you do not publish as much, but there is
1073 a big difference between 1 or 2 papers a year and 0 papers a year. Be sure that the work
1074 from your PhD thesis is published.

1075 You do not need to focus on the highest profile, splashiest journals, but focus on
1076 the top disciplinary journals. I was the opponent for the PhD defense of Nic Jonzen at
1077 the University of Lund in 2001. While there, I learned that whenever a person had a
1078 paper accepted in *Science*, *PNAS* or *Nature* at Lund they celebrated with a bottle of
1079 champagne. But the same happened with papers in *The American Naturalist*, *Ecological*
1080 *Applications*, or *Ecology*.

1081 Continue to go to scientific meetings, even if you do not present your non-academic
1082 work. As in all things in life, networks matter and it is important that you maintain
1083 contacts.

1084 **Become Part of a Group – But Not Too Big of One**

1085 Caldera (2016) reviewed multi-authored papers in *Environmental Research Letters* and
1086 found that 61 of the top 100 papers had four authors or less (Figure 2). He concluded
1087 that that “small groups of authors often produce the work with the greatest impact,
1088 even in an inter-disciplinary setting. This suggests that it may be wise to institute policy
1089 changes that discourage inflation of author lists and that encourage the funding of research
1090 conducted by single investigators and small groups of researchers.”

1091 von Tunzelmann et al (2003) found that in many scientific fields per-capita productivity
1092 peaks at group size of 6-8 individuals, after which there is little per-capita gain. They
1093 noted that the threshold for the optimal productivity varies with discipline (e.g. lower in
1094 the humanities, arts, and pure mathematics and higher in clinical medicine).

1095 Other researchers have found that there is little relationship between the size of a
1096 research group and its per-capita output (Cohen 1980, 1981, 1991). For example, Cohen
1097 (1980, 1981) found in research groups ranging from 1 to about 50 scientists, the addition of
1098 one scientist to a group increased expected annual production by 1.1 papers and that a 0-
1099 truncated negative binomial distribution, which is the steady state frequency distribution
1100 for stochastic models of freely forming primate societies, fit the data very well. Cohen
1101 concluded that there is no reliable evidence for an optimal group size to maximize per
1102 capita research production.

1103 Thus, you have to find your own way in selecting group size. One helpful thing is to
1104 do the Myers-Briggs personality test (<https://www.myersbriggs.org/my-mbti-personality-type/mbti-basics/home.htm?bhcp=1>). Knowing your personality type provides a clue to
1105 the kind of group (and mentor) that will allow you to flourish. Also learn how authorship
1106 is assigned in the group (Duffy 2017) – for example, will the group leader on every paper
1107 as senior author even if his or her contribution is minimal and you have your own post-
1108 doctoral money (“but you were sitting in my lab”)? Finally learn how much time you
1109 will have to explore your own ideas versus working on those of the leader of the group.
1110

1111 **Ethical Issues in Funding**

1112 Here's a headline from the *New York Times* on 31 Dec 2016 "Scientists Loved and Loathed
1113 by an Agrochemical Giant" with subtitle "With corporate funding of research, 'There's
1114 no scientist who comes out of this unscathed.' "

1115 The article describes how the agrochemical company Syngenta funded research by
1116 James Cresswell at the University of Exeter to study the increasing rate of mortality of
1117 colonies of bees, and whether this was due to a the mite *Varroa destructor* or to pesticides
1118 (that is, natural or anthropogenic mortality). According to the article, Cresswell had
1119 been skeptical about the pesticides (perhaps why the company approached him), but his
1120 research kept pushing him away from the mites as the cause. The result of this was that
1121 the company pressed him to consider new data and a different approach.

1122 About these interactions, Cresswell noted that Syngenta had its own agenda, which
1123 was that he was supposed to show that the problem was the mite, not the chemicals. The
1124 article concludes "A review of Syngenta's strategy shows that Dr. Cresswell's experience
1125 fits in with practices used by American competitors like Monsanto and across the agro-
1126 chemical industry. Scientists deliver outcomes favorable to companies, while university
1127 research departments court corporate support. Universities and regulators sacrifice full
1128 autonomy by signing confidentiality agreements. And academics sometimes double as
1129 paid consultants."

1130 If you work in Pasteur's Quadrant, this ethical issue may arise and you need to be
1131 prepared before difficulties occur to face it. Remember that the closer an issue is to
1132 fundamental human goals and aspirations, the more difficult it is to separate scientific
1133 conclusions from other influences (Ludwig et al 2001).

1134 Similar questions arise in taking funding from the military or fossil fuel industries –
1135 not necessarily to work on problems of direct application to the military. For example, a
1136 considerable amount of the research concerning marine mammals and noise is funded by
1137 the US Office of Naval Research (and another large chunk by oil companies such as Shell
1138 and Mobile).

1139 It is incumbent upon each of us who works in Pasteur's Quadrant to wrestle with these
1140 issues and to determine a self-consistent approach for funding.

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