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Caging and uncaging genetics

Tom J. Little* & Nick Colegrave
University of Edinburgh
Institute of Evolutionary Biology
School of Biological Sciences
Edinburgh, United Kingdom

*tom.little@ed.ac.uk

15 **Abstract**

16 It is important for biology to understand if observations made in highly reductionist
17 laboratory settings generalise to harsh and noisy natural environments where genetic variation
18 is sorted to produce adaptation. But what do we learn by studying, in the laboratory, a
19 genetically diverse population that mirrors the wild? What is the best design for studying
20 genetic variation? When should we consider it at all? The right experimental approach
21 depends on what you want to know.

22

23 **Introduction**

24 Biologist of all kinds have the option of using defined genetic lines for experiments, for
25 example inbred lines of mice and rats, nematodes, fruit flies or *Arabidopsis* stocks. An
26 alternative is the use of outbred stocks, and, indeed, areas with especially strong traditions in
27 laboratory experimentation have been encouraged to pay greater attention to the study of wild
28 populations[1-3]. Along side this, we frequently hear calls for the expansion of experimental
29 designs to include more genotypes [4], or to place these genotypes in a wider range of
30 environments[5]. One advocated goal in all of this is to add an extra dose of realism to our
31 science, so to better understand how our treatments, or particular biological phenomena,
32 might manifest in a more natural context. There is potential to misinterpret this point.
33 Although calls for realism are typically heard within a particular context, the range of
34 contexts spans much of biology, for example immunology[6], toxicology[7, 8], host-parasite
35 interactions[9-11] or coevolution and the selective maintenance of breeding systems[12].
36 Thus, at stake are very general issues regarding why biologists choose a particular
37 experimental design, and the question becomes: what is gained and lost in a laboratory study
38 of, at one extreme, an outbred population compared to, at the other extreme, the study of a
39 single genotype? Here, we synthesize these ideas, highlighting the pros and cons of the
40 various ways experiments can incorporate genetic variation.

41

42 **In Defense of One**

43 The essence of an experiment is reduction in the number of explanatory variables.
44 Experimental control of environments leads to less variation within treatments, improving our

45 chances of finding differences between treatments. As a simple example, we would not want
46 to study the effect of a treatment on plants held in a faulty incubator whose lights flickered on
47 and off randomly throughout the day and whose temperature varied spatially inside. A
48 properly functioning incubator would eliminate photoperiod as a source of variation, and limit
49 environmental heterogeneity amongst our experimental plants, improving the power of our
50 experiment.

51 Thus, in many cases, if the aim of a study is to test the effect of a treatment, then the most
52 effective approach is to make use of organisms that are as genetically uniform as is
53 practicable (Fig 1A). A single inbred line, or an ameiotic clone, is ideal. The reasoning behind
54 this is simple: in any experiment we are trying to separate variation caused by our treatment
55 from variation due to other sources. If we carry out a study on genetically diverse subjects,
56 then genetic variation for the character of interest, just like a faulty incubator, can make it
57 more difficult for us to detect the treatment effects that we are interested in. Working with a
58 single genotype removes this source of variation, maximising our power to detect effects of
59 treatments. Thus, the argument for carrying out experimental studies on a single genotype is
60 simply a subset of the argument for using controlled laboratory conditions: we wish to reduce
61 variation from sources that we are not currently interested in. We also often have the
62 opportunity to choose genotypes. For example, if doing a behavioural experiment, one would
63 not choose a strain for which the behaviour of interest had been bred out. Adding genotypes,
64 some of which show a different expression of the trait, can reduce our capacity to study the
65 trait of interest.

66 The success of this approach cannot be understated. The nematode *Caenorhabditis elegans*
67 was initially isolated from a single source, and thus essentially a single genotype. The vast
68 majority of laboratory *Escherichia coli* are descendants of one of two isolates. For decades,
69 in the absence of noise-making genetic variation, laboratory models have contributed
70 immensely to our understanding of developmental biology, signaling pathways, meiosis,
71 ageing, adaptation and phenotypic plasticity, amongst many other important biological
72 phenomena. One reason this has been achieved, aside from the powerful features and
73 malleability of these organisms, is that defined lines can be shared, and studies replicated in
74 different laboratories. There is no question genetically uniform model organisms have proven
75 generalizable to other environments, other genotypes of the same species, and indeed across
76 large phylogenetic distances.

78 **Genetic Variation and Outbred Populations**

79 Model organisms have also contributed to our understanding of the effects of genotype on
80 phenotype, in particular where we have crafted lines that differ at only a single locus. In these
81 cases, we study genetic variation, but in the simplest possible format in the tradition of
82 experimental biology. Expanding to more genetic variation might involve incorporating a set
83 genotypes initially isolated, for example, from a wild population but then maintained as
84 separate, often inbred and uniform, stocks in the lab. At the farthest end of the spectrum, we
85 might even perform our experiment on an outbred population, which will contain much
86 uncharacterised genetic diversity. The latter fosters understanding of how wild populations
87 might respond to a treatment, yet comes with important limitations. At the very least, if this
88 genetic variation affects the character that we are interested in, then its presence in our study
89 will add noise, making any treatment effects more difficult to detect. And even if issues of
90 reduced power can be minimized by increased sampling, if different genotypes respond to the
91 experimental treatment in different ways (there is a genotype by treatment interaction), the
92 average effect of a treatment on a diverse uncharacterized population may lead to misleading
93 conclusions. Consider, in the extreme, a particular treatment that has an effect on the
94 character that you are interested in, but in half of the genotypes in your population the effect
95 is positive, whilst in the other half the effect is negative by the same amount (Fig 1B). An
96 experiment on a diverse population will see no overall effect of the treatment. In contrast, if
97 we carry out the experiment a set of genetically uniform lines, strains or clones, we will
98 discover strong effects that vary amongst genotypes (Fig 1C). By essentially repeating our
99 experiment on this set of genetically uniform lines (Fig 1C), not only do we benefit from the
100 increased statistical power that comes with reduced variation (compared to an outbred
101 population), a comparison of treatment effects among lines gives us a measure of how
102 variable any treatment effects will be in a genetically diverse population. When
103 experimenting on an outbred population, we can say nothing about how treatment effects vary
104 with genotype (GxE), unless we have pedigree information.

105 GxE interactions are common, and they can be viewed as an argument against the use of
106 single genotypes in experimental work[4, 13]. The conclusions of single genotype studies are
107 limited to that specific genotype, so the argument goes, and it is only by experimenting on
108 genetically diverse material that we can generalise to the diverse populations that we care

109 about. However such an argument confuses the population to which the statistical conclusions
110 can be formally applied to (which is indeed the genotype), with the population to which the
111 biological conclusions might be applied to, which might be the genotype, or the population it
112 came from, or the species and beyond depending on the biological trait we are concerned
113 with. And of course, concerns about the range of conditions over which our conclusions
114 might apply are not restricted to genetic effects. The experiment might also not generalise to
115 temperatures other than those studied, to different food types, another lab, or to a different
116 country or planet. The statistical conclusions of single environment studies are limited to that
117 specific environment, or the specific range of treatments studied. For example an experiment
118 that compared phenotypes at 10°C and 20°C can, technically, only draw conclusions about
119 these specific temperatures, yet we are often comfortable generalising such results to draw
120 conclusions about the consequences of living at lower versus higher temperatures[14].

121 Thus, we need to carefully consider what we are trying to understand with an experiment, and
122 what we might be trying to generalise *to* when we add genotypes to our experiment. By
123 adding many genotypes to our experiment, in addition to the obvious consequence that we can
124 evaluate genetic variation (if we replicate genotypes), we also may gain a better
125 understanding of the average response of a population. Understanding population average
126 responses may be the goal of some ecologists, epidemiologists or evolutionary biologists. For
127 example, we might have an interest in the keystone species of an ecosystem we harvest from,
128 or in particular pathogens we want to control. In many other cases, however, we are not
129 interested in our study species - our study species is a model for a trait of interest. If the extra
130 genotypes merely get us closer to understanding our study species, we are not substantially
131 closer to generalising beyond our species because variation between species is often much
132 larger than variation within species. It is intriguing that many scientists are prepared to accept
133 mice as a model for human immunity (i.e. to generalise across species, from mice to humans,
134 in research of immense health and welfare importance), yet would quibble about where
135 studies on one genotype might not generalise to other genotypes of the same species. Imagine
136 that work on the *Caenorhabditis elegans* N2 strain had been rejected by the scientific
137 community for fear its patterns would be limited to that strain. Similarly, Lenski's long-term
138 study of bacterial evolution began with a single bacterial cell – an elegant design choice from
139 which we gladly draw conclusions we hope will apply to other bacteria and far beyond.

140

141 **Concluding remarks**

142 When creating laboratory lines, a few issues need to be considered. First, although the study
143 of defined lines fosters comparisons of results from different labs, and thus extensive
144 repetition of experiments, the maintenance of the original ‘wild-type’ in different laboratories
145 can generate significant divergence between these ‘replicate’ wild-types, as well as between
146 laboratory strains and new field isolates. Second, the process of establishing lines in the
147 laboratory often involves inbreeding, and inbred lines may suffer increased developmental
148 instability, which is not a source of variation we are likely to be interested in. Moreover,
149 inbreeding generates homozygosity across much of the genome, which does not capture
150 certain genetic effects that wild genotypes almost invariably experience. On the flip side,
151 inbreeding reveals the phenotypic effect of recessive mutations, which has been a
152 tremendously powerful tool in biology.

153 Establishing lines from wild stock may also involve selection. A salient example concerns the
154 effect of caloric restriction on ageing. Caloric restriction has been shown to lead to longer life,
155 but this observation appears to be more prevalent in long-term laboratory lines than in wild
156 populations. It has been speculated that some model organisms have been selected for
157 enhanced fecundity, which trades-off with longevity, and caloric restriction has a much larger
158 effect in this situation than in wild lines that have not been strongly selected for early
159 reproduction[15]. Some laboratory animals have probably also been selected for depressed
160 immune responsiveness, which reduces problems of autoimmunity in low infection laboratory
161 environments [3], but compromises our ability to understand natural immune responses.

162 Thus a question that arises in laboratory comparisons of how different genotypes respond to a
163 treatment is ‘does a potentially unusual set of laboratory genotypes tell us more about the
164 genetic variation basis of performance than an outbred population?’ This issue is similar to
165 the generalisation concern for single clone experiments. Laboratory populations inevitably
166 misrepresent their wild friends to some degree (analogously, this issue burdens the
167 establishment of stabilised cell lines). The worry that our laboratory organisms are a little
168 odd, this must be balanced against the benefits of choosing the most powerful design for the
169 question we have. We suggest that confounding variables - inbreeding depression or
170 laboratory selection - are topics for valuable, but separate, downstream studies. This should be
171 acceptable to us, if we are prepared to use mice as a model for the human immune system, or
172 if we are encouraged by the flood of generalisable biological data that organisms such as *C.*

173 *elegans* or *Drosophila* have provided.

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220 Figure 1. A hypothetical outbred population of 24 genotypes to be sampled and studied in the
221 laboratory to determine the effect of a particular treatment, for example exposure to a heat
222 shock. Red represents the heat-shock treatment and blue is the control. Half of the genotypes
223 can be expected to respond positively to the treatment by growing a thicker carapace (thick
224 lines), while the other half are expected to respond by growing a thinner carapace (dashes).
225 Here are three possible study designs:

226 **A)** Replicated single genotype studies have the most power to detect a treatment effect, albeit
227 only in one direction. If, for example, we are primarily interested in the biological
228 consequences of a thicker carapace, then (A) is the optimal study design.

229 **B)** A random sample of the outbred population can not detect genetic effects, and may not
230 even detect a treatment effect if the negative trait values are similar in magnitude to the
231 positive ones. Increasing the number of genotypes will not help. The experiment represented
232 by (B) does a good job of representing the population average effect, though this may be of
233 little value if the research is interested in carapace thickening in response to heat shock.

234 **C)** A set of eight replicated inbred lines can reveal genetic variation and detect a treatment
235 effect by revealing treatment by genotype effects.

