

# 1 **Synthetic biology as an empirical tool for evolutionary theory**

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## 14 **Abstract**

15 Evolutionary biology has traditionally inferred process from patterns in extant organisms and the  
16 fossil record, leaving many foundational questions constrained by their historical nature. Over the  
17 past two decades, synthetic and high-throughput approaches — including deep mutational  
18 scanning, genome editing, ancestral sequence reconstruction, engineered mutators, and random-  
19 sequence assays — have made it possible to test these questions directly by constructing,  
20 perturbing, and replaying evolutionary systems. Here, we review how these approaches reshape  
21 several foundational questions: the distribution of mutational effects, the structure and navigability  
22 of fitness landscapes, the evolution of evolvability, developmental constraint, historical  
23 contingency, and the engineering of evolutionary systems. Across these domains, synthetic  
24 experiments are exposing unexpected mechanistic detail that refines and extends classical theory  
25 — revealing, for example, how strongly mutational effects depend on environmental and genetic  
26 context, how ruggedness can coexist with broad accessibility on fitness landscapes, how genotype–  
27 phenotype maps are intrinsically biased and heterogeneous, and how random sequences carry  
28 latent functional potential that may serve as raw material for later innovation. As these  
29 technologies continue to expand the empirical reach of evolutionary biology, theory in turn  
30 sharpens the questions they are best suited to address — an iterative dialogue between experiment  
31 and theory that brings us closer to understanding how life evolves.

32

## 33 **Significance Statement:**

34 Evolutionary biology has long faced a fundamental limitation: evolution unfolds over deep time,  
35 and we can study only the organisms and fossils that happen to have survived — leaving many  
36 central ideas about how evolution works difficult to test directly. New laboratory tools now allow

37 scientists to build, modify, and replay evolutionary systems from scratch, providing direct  
38 experimental tests of these ideas. This review shows that such experiments are refining — and  
39 sometimes overturning — long-held views, for example by revealing that mutational effects  
40 depend strongly on context, that many evolutionary trajectories are shaped by historical  
41 contingencies, and that even random DNA can carry biological function. Together, these advances  
42 shift the field from observing evolution to experimentally probing its underlying molecular  
43 mechanisms.

44 **Keywords:** Evolutionary Theory; Molecular Evolution; Synthetic Biology; Adaptive  
45 Landscapes; Historical Contingencies; Ancestral Sequence Reconstruction

## 46 1. Introduction

47 Evolutionary biology has long been characterized by a productive interplay between theory and  
48 empirical observation, with each continually informing and refining the other. Early empirical  
49 challenges to transformist ideas, such as comparisons of ancient and extant organisms,  
50 foreshadowed the more systematic synthesis achieved by Charles Darwin in *On the Origin of*  
51 *Species*<sup>1</sup>, where conceptual reasoning was tightly interwoven with diverse lines of evidence.  
52 Subsequent discoveries, notably the fossil of *Archaeopteryx*, reinforced evolutionary predictions  
53 by revealing transitional forms. Over time, major theoretical advances have provided unifying  
54 frameworks: the foundations of population genetics laid by Ronald A. Fisher<sup>2</sup>, J. B. S. Haldane<sup>3,4</sup>,  
55 and Sewall Wright<sup>5,6</sup> formalized the roles of mutation, selection, drift, and migration, while later  
56 developments—including the neutral theory advanced by Motoo Kimura<sup>7–9</sup> and Tomoko Ohta<sup>10</sup>,  
57 as well as models of adaptive dynamics and evolutionary game theory—extended evolutionary  
58 analysis into new domains. In parallel, empirical support has expanded across genetics,  
59 paleontology, comparative biology, biogeography, and controlled experiments, from early work  
60 in *Drosophila* to long-term microbial evolution studies. Together, these intertwined empirical and  
61 theoretical traditions have transformed evolutionary biology into a quantitatively grounded  
62 discipline capable of explaining and predicting patterns of biological diversity.

63 Yet the sources of empirical information have been limited by the inescapable fact that evolution  
64 is ultimately a historical process. Thus, many of the questions in evolutionary theory have been  
65 constrained by inferring what processes have underlain the observations of “design features of  
66 ‘finished products’”<sup>11</sup>. The study of evolutionary processes has been greatly improved by the  
67 capacity to run evolutionary simulations using computational programmes<sup>11–14</sup>. Though useful to  
68 investigate dynamical processes and to act as proof-of-principle of potential evolutionary avenues,  
69 such simulations were still restricted to stark simplifications of biological parameters. Over the  
70 last few decades, however, new molecular and synthetic approaches have provided complementary  
71 experimental tools that allow us to revisit foundational ideas directly in the laboratory. Techniques  
72 such as high-throughput DNA sequencing, genome editing, and synthetic gene construction now  
73 enable researchers to directly test classical evolutionary hypotheses, often affirming their  
74 relevance while also challenging long-held assumptions<sup>15–17</sup>.

75 These approaches, which we will broadly refer to as “synthetic approaches”, as opposed to the  
76 “traditional” empirical evidence based on sampling natural variation, have become a strong  
77 experimental tool that has allowed us to empirically fill the observational gaps produced by  
78 extinction, and to explore the space of possibilities beyond the realm biological evolution has

79 explored over the eons that precede us. In an analogous way to comparative morphology, sequence  
80 comparisons between different species allow identification of possible changes that relate to  
81 phenotypic changes that lineages followed. But the power of synthetic biology is that, by  
82 reconstructing the mutational steps that a lineage took to evolve a novel trait in the context of other  
83 organisms, it is possible to assay the extent at which a genetic change may have caused phenotypic  
84 traits. For example, synthetic biology has been useful to examine the genetic causes of the  
85 limbllessness of snakes<sup>18</sup>, the eyespots of butterfly wings<sup>19</sup>, the chiroptagium of bats<sup>20</sup>, the wing-  
86 like fins of flying fish<sup>21</sup>, or the exaggeration of sexually selected traits like the boxing first pair of  
87 legs of *Drosophila prolongata*<sup>22</sup>.

88 More broadly, synthetic biology is propelling evolutionary research into new empirical territory,  
89 not only by dissecting the mechanistic basis of morphological and physiological change but also  
90 by enabling direct tests of the rules governing evolutionary processes. In this way, it has become  
91 a powerful tool for advancing evolutionary theory. In this review, we revisit classical questions in  
92 the field and examine how cutting-edge experimental approaches are providing fresh insights into  
93 them. Focusing on six major questions in evolutionary theory, we highlight representative  
94 examples in which synthetic methods bridge theoretical concepts and empirical data and  
95 emphasize how these emerging tools are reshaping our understanding of evolutionary dynamics,  
96 concluding with a forward-looking perspective on open questions and future directions.

## 97 **2. Foundational questions revisited through synthetic approaches**

### 98 **2.1. The interplay between mutations and selection**

99 Genetic mutations play a central role in evolution by providing the ultimate source of heritable  
100 variation upon which natural selection and other evolutionary forces act. Despite this fundamental  
101 role, the evolutionary importance of mutations themselves—their rates, spectra, and biases—has  
102 been a recurrent point of debate. During the early twentieth century, mutationist perspectives  
103 gained prominence during the so-called “eclipse of Darwinism”<sup>23</sup>, emphasizing the creative role  
104 of mutations in shaping evolutionary change. These views were later largely relegated by a  
105 selectionist framework, grounded in the assumptions that individual mutational effects are  
106 typically small relative to the strength of selection, and that populations harbor sufficient standing  
107 genetic variation to respond rapidly to environmental change. The elucidation of the structure of  
108 DNA and the genetic code transformed our understanding of the nature of mutations and their  
109 evolutionary consequences, leading to major conceptual advances such as the neutral theory of  
110 molecular evolution, renewed interest in gene duplication as a driver of novelty, and the rise of  
111 comparative and population genomics. These molecular advances have promoted the idea of  
112 “mutation-driven evolution”<sup>24</sup>.

113 The ability to study mutations in a synthetic and highly controlled manner has further highlighted  
114 the evolutionary relevance of the mutational process itself. High-throughput generation of large  
115 numbers of sequence variants, their efficient introduction into biological systems, and precise  
116 manipulation of the molecular machinery that governs mutation rates and spectra now allow direct  
117 experimental interrogation of how mutational processes shape evolutionary trajectories. In  
118 combination with experimental evolution approaches, artificial genetic constructs and engineered

119 mutational regimes make it possible to explore evolutionary dynamics under conditions that both  
120 recapitulate and extend beyond those encountered in natural systems (**Figure 1**).

### 121 **The distribution of fitness effects of mutations is dynamic**

122 The relative importance of selection and mutational input on the potential of affecting the  
123 evolutionary process depends primarily on (i) the rate at which mutations occur, and (ii) the  
124 phenotypic and fitness effects of individual mutations. Our capacity to estimate both of these  
125 parameters has substantially improved over the last decades. In the case of the effects of mutations  
126 (for an overview of advances in mutation rate estimations see Padilla-Iglesias and Majic 2026<sup>25</sup>),  
127 synthetic biology has been fundamental. Early efforts to estimate the distribution of fitness effects  
128 of mutations (DFE) relied on indirect approaches, most prominently mutation–accumulation  
129 experiments in *Drosophila*<sup>26,27</sup>, bacteria<sup>28</sup>, and yeast<sup>29</sup>. By relaxing natural selection and allowing  
130 mutations to accumulate, these studies quantified the resulting declines in fitness and inferred that  
131 most new mutations are deleterious, with a wide distribution of effects. Parallel insights came from  
132 comparative genomics, where patterns of synonymous and nonsynonymous substitutions were  
133 interpreted through the lens of the neutral and nearly neutral theories<sup>7,10,30,31</sup>. Together, these  
134 classical approaches established the prevailing view that the DFE is skewed toward neutrality and  
135 deleteriousness, with beneficial mutations being rare—even though the precise shape of the  
136 distribution remained difficult to resolve.

137 Synthetic biology techniques have added much-needed nuance to classical interpretations of  
138 mutational effects. In particular, the advent of deep mutational scanning has enabled direct, large-  
139 scale measurement of the fitness consequences of hundreds of thousands of specific gene variants  
140 in parallel. These studies challenge the simplistic view that most mutations are strongly  
141 deleterious, instead revealing different properties of DFEs such as: (i) most single mutations are  
142 neutral or slightly deleterious, with only a small fraction being beneficial<sup>32,33</sup>; (ii) the shape of the  
143 DFE depends strongly on environmental context<sup>34,35</sup>; (iii) saturating proteins with mutations has  
144 revealed pervasive epistasis, whereby the effect of a mutation depends on the presence of others  
145<sup>36,37</sup>; (iv) many DFEs are well described by truncated exponential distributions, such that small  
146 shifts in the mean markedly affect the probability of rare beneficial variants, highlighting the  
147 importance of the distribution’s “tail” for adaptation<sup>30,38</sup>; and (v) the mean effect of new mutations  
148 often declines with increasing background fitness, suggesting that physiological and  
149 environmental constraints modulate the DFE<sup>39–41</sup>.

150 Together, these results illustrate that DFEs are not static properties of mutations alone, but dynamic  
151 features shaped by molecular context, environment, and evolutionary history.

### 152 **The fitness effect of synonymous mutations**

153 The classical neutral theory proposed by Motoo Kimura treated most mutations at the molecular  
154 level as selectively neutral and regarded synonymous substitutions in particular as a paradigmatic  
155 example of neutral variation, given that they do not alter amino acid sequences<sup>7,8</sup>. Synonymous  
156 changes, however, are now recognized as potentially having substantial functional and fitness  
157 consequences<sup>33,42–47</sup>, by altering transcription, mRNA splicing, stability, and secondary structure,  
158 as well as translation efficiency and co-translational protein folding<sup>42</sup>. However, recent evidence  
159 based on high-throughput experiments has been controversial: in yeast, many synonymous variants  
160 were found strongly non-neutral<sup>48,49</sup>, although technical concerns have been raised with the

161 finding<sup>50</sup>. Similar results have been observed in bacteria, where synonymous edits in *E. coli* can  
162 impair growth in a condition-dependent manner<sup>45</sup>. Beyond deleterious outcomes, synonymous  
163 mutations have also been implicated in adaptation, for example in experimentally evolved  
164 *Pseudomonas fluorescens* populations<sup>44</sup>. However, high-throughput genome-editing experiments  
165 in human cells supported the traditional view that synonymous mutations rarely have fitness  
166 effects<sup>51</sup>. Therefore, a conclusive understanding about the neutrality of synonymous mutations  
167 awaits more empirical data.

## 168 **Mutational biases may shape evolutionary trajectories**

169 Classical population genetics often assumed mutation to be isotropic, with equal probability of  
170 change in all directions of sequence space, but this view has been overturned by accumulating  
171 evidence that mutation processes are inherently biased (e.g., transition–transversion asymmetries  
172 and pervasive GC→AT biases). While it was traditionally argued that natural selection would  
173 dominate over such biases except under neutrality, recent theoretical and experimental work shows  
174 that mutation bias can influence the trajectory of adaptation by shaping the pool of available  
175 variants. Some of this evidence comes from experiments focused on synthetic biology approaches.  
176 A clear experimental demonstration comes from work by Deepa Agashe and colleagues in  
177 *Escherichia coli*, who engineered strains with altered mutation spectra by modifying DNA  
178 replication machinery<sup>52</sup>. By effectively rotating the direction of mutational bias, they showed that  
179 the supply and distribution of beneficial mutations depend strongly on mutation spectrum. Strains  
180 biased toward mutation types rare in the wild type accessed a broader set of beneficial mutations,  
181 whereas those exaggerating the native bias rapidly depleted accessible beneficial variation,  
182 exhibiting up to an order-of-magnitude reduction in beneficial mutations in the distribution of  
183 fitness effects<sup>52</sup>. These results provide direct empirical support for the prediction that shifts in  
184 mutation bias can redirect adaptive evolution by altering which mutational paths are available,  
185 allowing populations with novel biases to access adaptive solutions that might otherwise remain  
186 unexplored.

## 187 **Gene duplication as a source of innovation**

188 Gene duplication has long been considered a central engine of evolutionary innovation. Susumu  
189 Ohno proposed that duplication frees one gene copy from selective constraint, allowing it to  
190 accumulate mutations that may eventually confer a new function (“neofunctionalization”), while  
191 the other copy preserves the ancestral role<sup>53</sup>. Synthetic approaches have begun to test this scenario  
192 directly. The Schaerli group<sup>54</sup> constructed *E. coli* strains carrying either one or two copies of a  
193 fluorescent protein gene and evolved them under selection for altered color output. Two-copy  
194 populations indeed accumulated more mutations and tolerated greater genetic diversity, consistent  
195 with relaxed constraint, yet they did not consistently evolve the target trait faster than single-copy  
196 strains. Often, one duplicate decayed into a pseudogene before it could be repurposed, illustrating  
197 that duplication enables but does not guarantee innovation. These findings nuance Ohno’s model  
198 by showing that the fate of duplicates depends on additional factors—selection regime, regulatory  
199 context, and the chance order of mutations.

200 A complementary framework, the escape from adaptive conflict (EAC) model, proposes that an  
201 ancestral gene performing multiple functions can be constrained by trade-offs, and that duplication  
202 allows each copy to specialize and resolve this conflict<sup>55,56</sup>. Support for EAC comes largely from  
203 ancestral reconstruction studies (see Section 2.5), which show how combining forward evolution

204 and historical approaches refines classical views of innovation. Classical models further assume  
205 that loss-of-function mutations in one copy are neutral. However, a recent study has shown that  
206 when duplicated genes interact, a defective copy can impair its partner, meaning purifying  
207 selection can maintain both copies longer than expected. For example, a deep mutational scan of  
208 a tetrameric protein found that ~6% of harmful mutations disrupt the second copy, helping preserve  
209 redundancy over evolutionary time<sup>57</sup>.

## 210 **2.2. The navigability of fitness landscapes**

211 Theoretical efforts to understand adaptation have long relied on spatial metaphors that map  
212 evolutionary change onto abstract representations of fitness. Sewall Wright’s fitness landscape  
213 provided an intuitive framework in which populations move across adaptive peaks and valleys  
214 under the combined influences of mutation, selection, drift, and recombination<sup>5,6</sup>. Complementing  
215 this view, Fisher’s geometric model of adaptation formalized how mutations of varying effect sizes  
216 influence fitness in high-dimensional phenotypic space, predicting that small-effect mutations are  
217 more likely to be beneficial as populations approach an adaptive optimum<sup>2</sup>. Subsequent theoretical  
218 work explored how the topology of fitness landscapes influences evolutionary dynamics. Models  
219 of rugged fitness landscapes, including those developed by Kauffman, predicted that increasing  
220 epistasis and landscape ruggedness can constrain adaptive walks, trap populations on local optima,  
221 and generate complex evolutionary trajectories<sup>58–61</sup>. The fitness landscape metaphor treats  
222 evolution as populations moving on a high-dimensional map, where the location corresponds to  
223 genotype and the elevation corresponds to fitness<sup>2,5,6</sup>. A central question is ruggedness—the  
224 prevalence of local peaks (genotypes whose single-mutational neighbors are all lower in fitness)  
225 and valleys (steps that reduce fitness)—and whether ruggedness traps populations on suboptimal  
226 solutions<sup>58–61</sup>. Classic theory debated how populations could traverse valleys.

227 Advances in synthetic biology have enabled direct empirical mapping of fitness landscape  
228 topology at unprecedented scale. For example, an exhaustive study of the antibiotic resistance  
229 enzyme dihydrofolate reductase in *Escherichia coli* quantified the fitness of over 260,000  
230 genotypes using genome editing and competitive assays, revealing a highly rugged landscape with  
231 numerous local peaks<sup>62</sup> (**Figure 2A**). Despite this extensive epistasis, the global fitness optimum  
232 remained broadly accessible, with many mutational trajectories proceeding uphill without  
233 traversing deep fitness valleys<sup>62</sup>. Similar “rugged yet navigable” architectures have been observed  
234 in regulatory DNA, including transcription factor binding sites mapped via massively parallel  
235 reporter assays (**Figure 2B**), where compensatory changes in core motifs, flanking regions, or  
236 spacing create ridge-like paths that maintain function<sup>63–65</sup> (**Figure 2C**). Other high-throughput  
237 studies across fluorescent proteins<sup>32</sup>, enzymes<sup>33,66</sup>, regulatory elements<sup>67,68</sup>, and viral proteins<sup>69,70</sup>  
238 likewise report pervasive epistasis alongside neutral networks<sup>8,31,71</sup> and multiple accessible  
239 optima<sup>72,73</sup>. Together, these findings suggest that while biological fitness landscapes are complex,  
240 their structure often permits evolutionary exploration through “multidimensional bypasses”<sup>74</sup>, and  
241 connected ridges and plateaus, mitigating the constraining effects of local peaks. Rather than  
242 trapping populations, ruggedness may instead reduce predictability, as stochastic mutational paths  
243 can lead different populations to distinct adaptive peaks.

244 High-throughput fitness assays in microbes have also enabled systematic characterization of  
245 genotype-by-environment (G×E) interactions, revealing dynamic “fitness seascapes”<sup>75–79</sup> (**Figure**  
246 **2C**) in which peaks and ridges shift with external conditions such as drug concentration or

247 temperature. This is exemplified by high-throughput mutational studies of the molecular  
248 chaperone Hsp90 across diverse stressors (e.g., heat, salinity, oxidative stress)<sup>34,80,81</sup>, as well as by  
249 analyses of promoter and transcription factor binding site variation<sup>82</sup>. These studies show that both  
250 ruggedness and accessibility are environment-dependent: a given genotypic neighborhood may be  
251 highly constrained in one condition yet readily navigable in another. Accordingly, predictive  
252 power improves when models are formulated in joint GxE space, incorporating environment-  
253 dependent global epistasis. The environmental dependence of the distribution of fitness effects is  
254 thus mirrored at the landscape level, with context reshaping not only mutational effects but also  
255 the topography traversed during adaptation.

### 256 **2.3. The evolution of evolvability**

257 Dawkins asked whether evolvability can itself evolve<sup>12</sup>. Essentially, organismal features can be  
258 selected such that they pre-condition the range of adaptive solutions that become available.  
259 Dawkins referred to this as the “evolution of evolvability”—certain features act as “evolutionary  
260 watersheds” that can occasionally “open the floodgates to future evolution”.

### 261 **Cryptic variation**

262 The concept of cryptic genetic variation as an evolutionary resource gained molecular footing  
263 when Rutherford and Lindquist showed that the chaperone Hsp90 acts as a phenotypic capacitor  
264 in *Drosophila*: under stress, previously buffered genetic variation is released, producing diverse  
265 morphological phenotypes<sup>83</sup>. Similar capacitor behaviour has since been documented for  
266 chromatin regulators, prions, and DNA methylation<sup>84</sup>. But can cryptic variation actually accelerate  
267 adaptation? Recent experiments have directly tested the role of cryptic variation in evolution. In  
268 one study, researchers accumulated mutations in a gene encoding a fluorescent protein in *E. coli*  
269 under conditions where those mutations had no effect on the protein’s function (thus building up  
270 cryptic variation in the population). They then challenged these lineages to evolve a new  
271 function—shifting the protein’s fluorescence from yellow to green under strong selection<sup>85</sup>. The  
272 results were striking: populations that harbored cryptic genetic variation evolved novel adaptive  
273 phenotypes faster and reached higher fitness peaks than populations without that cryptic pre-  
274 variation<sup>85</sup>. The cryptically mutagenized populations explored more diverse genotypic solutions  
275 and ended up with distinct mutations, whereas the non-cryptic (unmutated) populations tended to  
276 converge on the same limited set of changes. In essence, cryptic variation “opened up” access to  
277 diverse adaptive peaks, allowing different replicates to discover otherwise inaccessible high-  
278 fitness genotypes. This empirically demonstrates a long-suspected principle: genetic variation that  
279 is initially neutral can set the stage for later evolutionary leaps when the environment shifts. It also  
280 showed that cryptic variation increases the contingency and unpredictability of evolution—  
281 different populations, even in identical conditions, took divergent paths and outcomes because  
282 they had different hidden mutations available to exploit<sup>85</sup>. This finding has deep theoretical  
283 implications. It affirms that the past evolutionary history (even neutral drift) can profoundly  
284 influence future adaptation (a form of evolutionary “memory”). It also provides a mechanistic  
285 basis for evolvability: robustness (tolerance of mutations) and adaptability are two sides of the  
286 same coin, where robustness allows accumulation of cryptic changes that later give adaptability  
287 when conditions change.

## 288 **Evolvability as an object of selection**

289 Evolvability itself can also be a direct target of selection. In a recent long-term experiment with  
290 the bacterium *Pseudomonas fluorescens*, Barnett et al.<sup>86</sup> (2025) forced populations to repeatedly  
291 switch back and forth between two ecological states, to select for populations with greater capacity  
292 to generate changes. Over time, one lineage evolved a special kind of “hypermutable” DNA  
293 sequence in a regulatory gene: a short repeat that made it very easy for mutations to occur. This  
294 repeat acted like a genetic switch, producing rapid and reversible changes that allowed the bacteria  
295 to adapt again and again. This experiment showed, for the first time in real time, that natural  
296 selection can favor genetic features that make a lineage more adaptable, proving that evolvability  
297 is not just a by-product of evolution but a trait that can itself evolve. In a related vein, experiments  
298 on protein evolvability have revealed how selection can shape future potential. One study<sup>74</sup>  
299 evolved yellow fluorescent protein under strong selection, then shifted to select a new color. The  
300 variants that emerged under strong selection were also more robust to mutations (better  
301 foldability), which allowed them to evolve the new function more quickly. Similarly, Bloom et  
302 al.<sup>87</sup> showed that proteins evolving in large populations tend to become more robust and stable,  
303 improving evolvability. These findings support the idea that selective regimes—including  
304 alternation of functional pressures or variation in selection strength—can shape not only present  
305 adaptation but also the potential for future adaptation.

## 306 **2.4. Developmental constraints**

307 The question of evolvability is further related to the topology of genotype–phenotype maps. For  
308 much of the 20th century, evolutionary theory treated variation as essentially random with respect  
309 to need—the Modern Synthesis emphasized natural selection as the creative force that shapes  
310 organisms, given raw material from mutations. However, not all phenotypic variation is equally  
311 likely: the way organisms are built (their developmental biology) can bias the variants that  
312 appear<sup>88,89</sup>. Developmental processes can favor some forms over others, not only limiting what is  
313 possible, but also facilitating the capacity to adapt and diversify<sup>90</sup>.

314 Synthetic biology tools have been useful to explore this question empirically at the molecular level.  
315 A long-standing argument holds that regulatory changes drive morphological evolution because  
316 they are more modular and less pleiotropic than coding mutations<sup>91–93</sup>, but recent evidence  
317 suggests that *cis*-regulatory mutations are often more pleiotropic than previously appreciated.  
318 Deep mutational scanning is beginning to resolve this debate: DMS of the yeast *TDH3* promoter  
319 shows that *cis*-regulatory mutations alter expression level and noise but cause milder fitness effects  
320 and lower pleiotropy than *trans*-regulatory mutations<sup>94,95</sup>. By contrast, DMS in developmental  
321 contexts—made tractable by high-throughput imaging—reveals that single point mutations in  
322 *Drosophila* enhancers can have large effects on the level and pattern of target gene expression<sup>96,97</sup>,  
323 challenging the assumption of low pleiotropy and highlighting the dense regulatory information  
324 packed into developmental enhancers.

325 To explore developmental constraint experimentally, deep mutational scans in developmental  
326 contexts become a useful tool. In terms of regulatory mutations—which are prime drivers of  
327 developmental evolution—it is possible to explore the effects of mutations using reporter  
328 constructs that become active in developmental contexts. Some examples using these systems have  
329 shown how developmental constraint manifests itself in highly pleiotropic enhancer mutations and

330 the difficulty in generating new patterns with a few mutations<sup>96-98</sup>. Random sequences may have  
331 high regulatory potential in *Drosophila* embryos, but their regulatory activity may be highly  
332 contingent on pre-established chromatin accessibility<sup>98</sup>. More empirical evidence is still needed  
333 for evaluating the ruggedness of the fitness landscape of developmental networks, but current  
334 evidence hints that it could be highly constrained. It would also be of interest to compare  
335 developmental networks to gene regulatory networks underlying other traits, e.g., metabolism or  
336 behavior, in order to understand whether high constraint is a unique property of development or a  
337 general feature of complex networks.

## 338 **2.5. The historical contingency of evolutionary change**

339 Despite their enduring influence, the power of selection as the driving force of trajectories over  
340 fitness landscapes has been debated. Gould and Lewontin famously questioned the adaptationist  
341 tendency to interpret biological traits as optimal solutions to selective problems, emphasizing  
342 instead the roles of historical contingency, developmental constraint, and correlated trait  
343 evolution<sup>99</sup>—all of which can affect the ruggedness, dimensionality, and accessibility of fitness  
344 landscapes and can, therefore, profoundly shape evolutionary outcomes, independent of selection  
345 strength alone. In the context of this criticism, Gould famously framed this issue through the  
346 metaphor of “replaying the tape of life,” asking whether evolution would retrace similar paths if  
347 history were reset, or whether chance events would lead to fundamentally different outcomes<sup>100</sup>.  
348 This question lies at the heart of debates about evolutionary determinism and about the extent to  
349 which selection, given similar starting conditions, leads to convergent solutions versus historically  
350 contingent trajectories<sup>101</sup>.

351 A common challenge in evolutionary biology is that we can observe only the present, not the  
352 past—yet evolution is inherently a historical process. An interesting approach to address these  
353 questions has come from our capacity to recreate the paths evolution took, but to also synthesize  
354 the paths it could have taken. Integrating this with experimental evolution and developmental and  
355 structural approaches offers a path toward understanding how contingency, constraint, and bias  
356 jointly determine the replayability—and predictability—of evolutionary change.

357 Ancestral sequence reconstruction (ASR) is a powerful molecular approach that seeks to infer  
358 ancient biomolecules (DNA, RNA, proteins) from extant sequences, synthesize them in the lab,  
359 and experimentally characterize their functions<sup>102,103</sup> (**Figure 2D**). In essence, ASR lets us “replay”  
360 evolution in retrospect: we can resurrect long-extinct proteins or gene sequences and examine how  
361 their properties differ from modern ones, thereby testing hypotheses about what changed and why.  
362 Among other interesting conclusions, findings from ASR have particularly informed about the role  
363 of contingency in molecular evolution.

364 Reconstructing a molecule’s evolutionary path allows direct observation of evolutionary  
365 constraint. By recreating ancient enzymes, scientists can introduce historical mutations one by one  
366 to see how function changes, revealing epistatic interactions that constrained certain evolutionary  
367 paths and channeled others. A famous example is the evolution of cortisol specificity in  
368 glucocorticoid receptors: ASR revealed that a particular mutation required to confer high  
369 specificity was only tolerated after another “permissive” mutation occurred, explaining why  
370 evolution took the path it did (historical contingency at the molecular level)<sup>104-107</sup>. These kinds of  
371 studies marry well with theoretical models of sequence space and adaptive walks, providing

372 concrete data on how many mutations, and in what order, are needed for new functions (**Figure**  
373 **2D**).

374 Ancestral reconstruction has also illuminated how gene duplication generates new function  
375 through the resolution of adaptive conflict (see Section 2.1). Reconstruction of an ancestral steroid  
376 hormone receptor before a duplication event showed that it bound multiple hormones with  
377 intermediate specificity, consistent with a single gene juggling multiple roles<sup>104</sup>. After duplication,  
378 one copy acquired degenerative mutations that abolished hormone-activated signaling but  
379 preserved DNA binding, converting it into a competitive repressor that fine-tunes the activity of  
380 its sister copy<sup>104</sup>. This division of labor resolved an apparent regulatory conflict and demonstrated  
381 that ostensibly deleterious loss-of-function mutations can drive innovation when they occur in a  
382 duplicated background<sup>104</sup>. Such studies provide direct experimental support for the escape from  
383 adaptive conflict model and show how synthetic resurrection complements forward-evolution  
384 experiments in dissecting the mechanistic origins of new functions.

385 Ancestral reconstruction has also allowed the mapping of ancient genotype–phenotype  
386 relationships. Herrera-Álvarez et al.<sup>108</sup> generated comprehensive libraries of ancestral transcription  
387 factors and their DNA response elements, assaying all possible combinations at historically  
388 variable sites. They found that the resulting genotype–phenotype maps were strongly biased: some  
389 phenotypes were far more likely to appear than others (anisotropy), and the set of phenotypes  
390 accessible from one genotype differed greatly from those around another (heterogeneity). These  
391 intrinsic biases steered ancient proteins toward the lineage-specific DNA-binding specificities that  
392 later evolved, showing that the structure of genotype–phenotype maps themselves can act as causal  
393 determinants of evolutionary outcomes. This work extends the fitness landscape framework by  
394 demonstrating that ruggedness and accessibility are not only properties of extant molecules, but  
395 also of the ancestral maps that shaped long-term evolutionary history.

396 Beyond informing molecular evolution, ASR can record ancient environmental conditions.  
397 Reconstructed enzymes from cyanobacteria and land plants show thermostabilities consistent with  
398 Earth's surface cooling from ~75 °C three billion years ago to ~35 °C by the Devonian, providing  
399 biophysical evidence that complements geological paleotemperature estimates<sup>109</sup>. Overall, ASR  
400 gives evolutionary biologists a remarkable time machine: instead of merely inferring the past, we  
401 can experiment on it. The insights from these resurrected molecules often uphold classical  
402 narratives (e.g., confirming a predicted ancestral function) but also add layers of nuance (e.g.,  
403 revealing previously unknown intermediate states or latent functions). By directly observing  
404 “missing links” at the molecular level, we enrich evolutionary theory with a more detailed  
405 understanding of how new biological functions and complexities have arisen through deep time.

### 406 **3. The evolving toolkit for engineering evolutionary systems**

407 Beyond high-throughput measurement of natural sequence variation, synthetic biology offers a  
408 fundamentally different empirical mode: building biological systems *de novo* and observing how  
409 evolution acts on them. Three classes of approach have proven especially productive.

410 Synthetic gene circuits and minimal genomes provide controlled platforms for evolutionary  
411 experimentation<sup>110,111</sup>. For example, the study of synthetic genetic circuits in *E. coli* has shown  
412 that certain gene regulatory network topologies are more evolvable towards specific functions than

413 others<sup>110,111</sup> (**Figure 2E**). By introducing engineered regulatory circuits or metabolic pathways into  
414 living cells, researchers can ask whether selection tunes their components toward predicted optima,  
415 thereby probing principles of adaptation in tractable systems<sup>112–114</sup>. The assembly of a minimal  
416 bacterial genome of approximately 473 genes (JCVI-syn3.0) provides a stripped-down chassis in  
417 which to ask whether selection regenerates discarded complexity<sup>115–118</sup>. Adaptive evolution on  
418 cells with such reduced genome showed that very few generations are needed for fitness to be  
419 improved (**Figure 2F**). On the eukaryotic side, the synthetic yeast genome project *Saccharomyces*  
420 *cerevisiae* 2.0<sup>119–121</sup> incorporates scrambleable genome segments designed to allow controlled  
421 rearrangement under selection<sup>109</sup>. Such systems test classical questions about the evolution of  
422 robustness, modularity, and genome architecture in regimes that natural genomes do not readily  
423 afford.

424 Engineered mutators decouple variation from natural mutation processes<sup>122</sup>. Programmable in vivo  
425 platforms such as EvolvR<sup>122,123</sup> can target hypermutation to specific loci using CRISPR-guided  
426 DNA polymerases, while systems such as TF-HighEvo<sup>124</sup> in *Drosophila* fuse a transcription factor  
427 to an activation-induced deaminase to concentrate germline mutations around its binding sites,  
428 raising local mutation rates roughly tenfold and generating heritable phenotypic diversity. These  
429 tools allow direct experimental control over the supply, location, and spectrum of variation,  
430 providing tests of how mutation rate and bias shape adaptation in regimes that natural mutagenesis  
431 cannot easily produce.

432 Random sequences interrogate the null expectation for biological function. What fraction of  
433 sequence space is functional, and how readily can selection convert noise into signal (**Figure 2G**)?  
434 Random RNAs and peptides expressed in *E. coli* often show measurable bioactivity, modulating  
435 growth in unexpected ways<sup>125</sup>. Random sequences also display surprisingly high structural  
436 propensity: ~40% of random 105-residue proteins adopt compact, protease-resistant  
437 conformations whether built from the full 20-amino-acid alphabet or from a reduced 10-residue  
438 prebiotic set<sup>126</sup>, and de novo designs using only the ten "early" amino acids can reconstruct ancient  
439 protein folds with thermostability and mutational robustness comparable to their full-alphabet  
440 counterparts<sup>127</sup>; the canonical alphabet itself appears to have been shaped by these foldability  
441 constraints, since prebiotically abundant non-canonical residues yield polypeptides that fold less  
442 reliably than the early canonical set<sup>128</sup>.

443 Random DNA sequences have also been used to study the potential of de novo evolution of gene  
444 regulatory elements. In bacteria, roughly 10% of random short DNA sequences placed upstream  
445 of a reporter in *E. coli* function as promoters without any evolution, and 60% become active after  
446 a single mutation<sup>129</sup>. Biophysical models calibrated on random mutant libraries reach similar  
447 conclusions, predicting that ~20% of arbitrary sequences drive measurable expression and that  
448 ~82% of inactive sequences lie one mutation away from a functional  $\sigma^{70}$  promoter<sup>130</sup>. This  
449 propensity for de novo regulatory emergence is even greater in random DNA than in extant  
450 genomic sequence: a recent massively parallel reporter assay found that promoters arise roughly  
451 threefold more readily from random DNA than from native *E. coli* sequences, because the genome  
452 contains fewer proto-binding sites for transcriptional activators than random DNA does<sup>131</sup>.  
453 Screens of approximately 100 million random promoters in yeast similarly show that random  
454 sequences frequently contain transcription factor binding motifs and span much of the natural  
455 expression range<sup>132,133</sup>. In *Drosophila* embryos, random sequences readily drive reporter  
456 expression at mid- and late-stage development but rarely in early embryos, where pre-established  
457 chromatin accessibility constrains regulatory activity<sup>98</sup>. Together, these results suggest that, in

458 many contexts, evolution must act as much to suppress unwanted activity as to create new  
459 function—and that natural genomes may even bear the signature of selection trimming away  
460 regulatory potential that random DNA possesses by default—reframing a longstanding question  
461 about the origins of regulatory novelty.

#### 462 **4. Conclusion**

463 Bringing together these threads, it is evident that we are in a golden age of empirical evolutionary  
464 biology that is deeply informed by and informing theory. The dialogue between past and future—  
465 between classical ideas and modern approaches—is enriching both. We find that many  
466 foundational concepts were essentially on the right track, but reality is often more quantitatively  
467 nuanced and interconnected than early models assumed. What do these advances provide for the  
468 future of evolutionary theory? The progression from measuring sequence–function relationships  
469 (**Figure 2A, B**) to characterizing fitness seascapes (**Figure 2C**) and reconstructing evolutionary  
470 history (**Figure 2D**), to testing architectural constraints and de novo function with synthetic  
471 systems (**Figure 2E–G**), to predicting evolutionary outcomes (**Figure 2H**) mirrors the broader  
472 intellectual trajectory of the field: from cataloguing evolutionary possibilities to forecasting which  
473 of those possibilities will be realized.

474 As empirical fitness landscapes and genotype–phenotype datasets expand, machine learning  
475 approaches are beginning to capture latent biophysical constraints and generalize across sequence  
476 space<sup>133,134</sup>, and deep neural models trained on exhaustive landscape data can infer fitness for  
477 unmeasured genotypes and guide experimental design<sup>135</sup>. Yet these advances also caution against  
478 overconfidence: replay experiments and large-scale maps consistently show that the order of  
479 mutations, clonal competition, and environmental fluctuations make evolution partly  
480 unpredictable<sup>136</sup>. A more realistic goal is probabilistic forecasting — estimating which outcomes  
481 are most likely and how probable different evolutionary paths are, rather than predicting single  
482 trajectories.

483 The horizon of tools keeps expanding. Long-read sequencing and ancient DNA may allow us to  
484 directly observe evolution over longer timescales (thousands of years) in natural populations,  
485 bridging the gap between lab time and geological time. CRISPR-based lineage tracking in  
486 multicellular organisms could open experimental evolution to plants and animals, not just  
487 microbes—one could envision tracking somatic evolution or virus evolution within a host in  
488 unprecedented detail<sup>137</sup>. These will pose new theoretical questions about heterogeneous  
489 environments and multilevel selection (e.g., cell-level vs. organism-level fitness). Synthetic  
490 biology also offers the possibility of constructing a minimal life form and seeing what directions  
491 evolution takes it in various worlds, akin to simulating alternative histories. Furthermore, as  
492 empirical fitness landscapes and genotype–phenotype datasets expand, the prospect of predictive  
493 evolution—anticipating likely adaptations in pathogens, tumors, or crops—moves from  
494 speculation toward practice.

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#### 509 **Competing interests**

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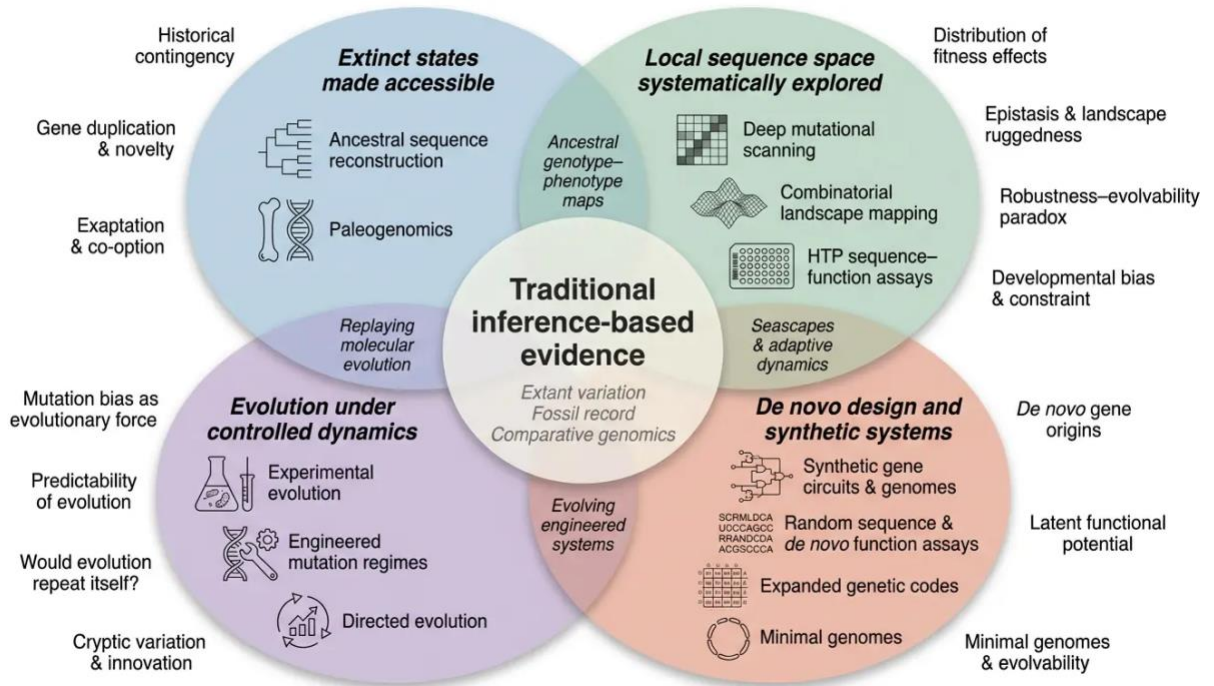
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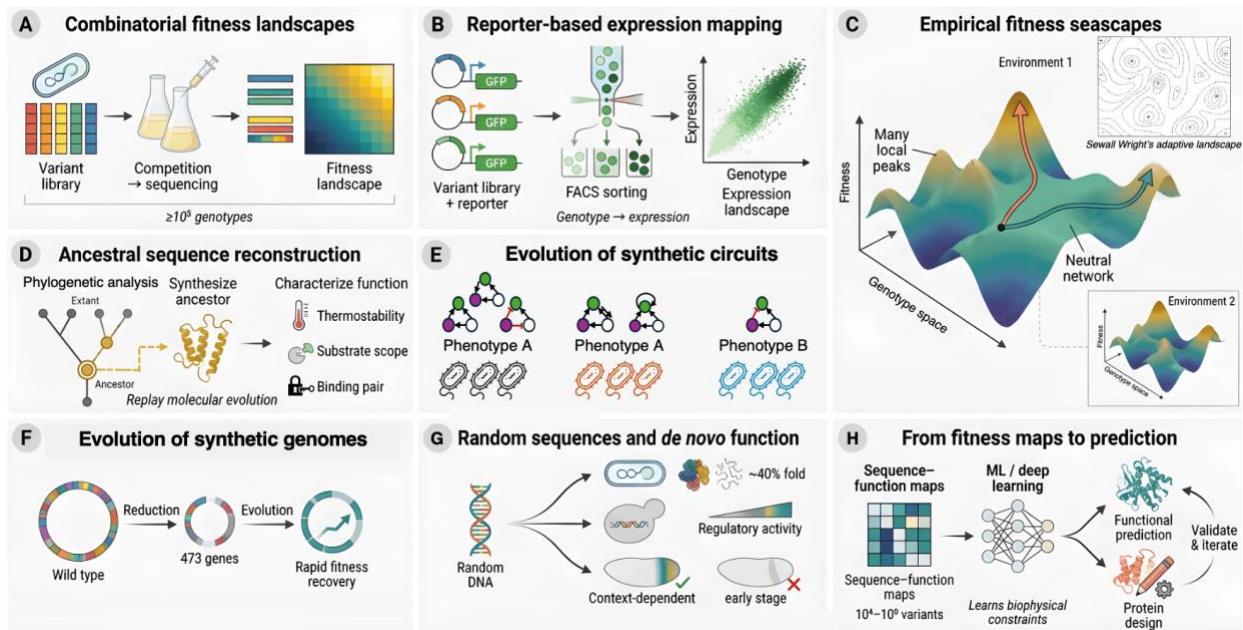


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820 **Figure 1. Synthetic and experimental approaches expand evolutionary access beyond natural observation.**

821 Evolutionary biology has traditionally relied on inference from extant variation, the fossil record, and comparative  
822 genomics (centre). Four complementary classes of approaches now extend this empirical reach. *Extinct states made*  
823 *accessible* (upper left): ancestral sequence reconstruction and paleogenomics resurrect ancient molecules and  
824 genotypes, enabling direct functional tests of evolutionary history. *Local sequence space systematically explored*  
825 (upper right): deep mutational scanning, combinatorial libraries, and high-throughput sequence–function assays map  
826 genotype–phenotype and fitness landscapes around natural sequences with unprecedented resolution. *Evolution under*  
827 *controlled dynamics* (lower left): experimental evolution, engineered mutation regimes, and directed evolution allow  
828 populations to be replayed, redirected, or exposed to defined selective pressures, testing how mutation bias,  
829 contingency, and repeatability shape adaptive trajectories. *De novo design and synthetic systems* (lower right):  
830 synthetic gene circuits, minimal genomes, expanded genetic codes, and random-sequence assays probe functional  
831 regions of sequence space that natural evolution may never have sampled, addressing what evolution could do rather  
832 than only what it did. Overlap zones highlight questions that are most effectively addressed by combining approaches:  
833 ancestral genotype–phenotype maps emerge from coupling sequence reconstruction with systematic landscape  
834 mapping; replaying molecular evolution integrates resurrected ancestors with forward experimental evolution;  
835 seascapes and adaptive dynamics link empirical landscape characterisation with controlled evolutionary experiments;  
836 de novo regulatory landscapes connect high-throughput assays with synthetic and random sequence libraries; and  
837 evolving engineered systems combines experimental evolution with synthetic circuits and minimal genomes. Classical  
838 evolutionary questions are shown at the periphery. Together, these approaches shift the field from observing  
839 evolutionary outcomes to experimentally testing evolutionary possibilities.

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**Figure 2. Experimental and synthetic approaches for studying evolution.** The figure is organized in three rows that follow a logical progression: the top row shows how high-throughput genotype–fitness maps are constructed and what the resulting landscapes look like, the middle row illustrates how evolutionary history is reconstructed and how synthetic systems test architectural constraints on evolvability, and the bottom row shows how synthetic genomes evolve, how random sequences reveal latent functional potential, and how empirical data feed predictive models. **A. Combinatorial fitness landscapes.** Large combinatorial variant libraries ( $\geq 10^5$  genotypes) are introduced into cells by genome editing and subjected to growth competition under defined selective conditions. Deep sequencing yields empirical fitness landscapes that quantify *in vivo* fitness for each genotype. **B. Reporter-based expression mapping.** Libraries of regulatory sequence variants are cloned upstream of a fluorescent reporter (e.g., GFP), expressed in cells, and sorted by fluorescence intensity using FACS. Sequencing of each sorted bin links sequence variation to quantitative regulatory output, producing high-resolution genotype–expression landscapes. **C. Empirical fitness seascapes.** In contrast to Wright's original (faint grey silhouette), empirical landscapes reveal richly structured topographies with many local peaks, ridges, and neutral networks. Two trajectories from a shared starting genotype illustrate that most peaks are reachable without crossing deep valleys: a direct adaptive path (coral arrow) climbs steeply to one peak, whereas neutral exploration (teal arrow) traverses a plateau before ascending a different peak. Crucially, landscape topography is environment-dependent: the same genotype space exhibits a different fitness surface under alternative conditions (inset, Environment 2), producing fitness "seascapes" in which peaks, valleys, and accessible paths shift with external context. **D. Ancestral sequence reconstruction.** Phylogenetic analysis of extant sequences is used to infer ancestral states, which are then synthesized in the laboratory and experimentally characterized for properties such as thermostability, substrate scope, and binding specificity — effectively replaying molecular evolution. **E. Evolution of synthetic circuits.** Synthetic gene regulatory networks with distinct topologies are introduced into cells and evolved under selection. Different circuits producing the same phenotype (Phenotype A) can employ different underlying architectures, while alternative wiring can produce a different phenotype (Phenotype B), demonstrating that network topology constrains which evolutionary outcomes are accessible. **F. Evolution of synthetic genomes.** A wild-type bacterial genome is radically reduced to a minimal set of ~473 essential genes, then subjected to experimental evolution. Rapid fitness recovery demonstrates that even highly simplified genomes retain substantial evolvability and can reinvent functions that were stripped during reduction. **G. Random sequences and de novo function.** Random DNA sequences, never exposed to natural selection, are tested for biological activity across three systems. In *E. coli*, ~40% of random polypeptides adopt compact, folded conformations. In yeast, random sequences frequently contain transcription factor binding sites that drive measurable regulatory activity. In *Drosophila* embryos, random sequences can drive reporter expression at late developmental stages but not at early stages, revealing that functional potential is widespread but context-dependent. **H. From fitness maps to prediction.** High-throughput sequence–function maps encompassing  $10^4$ – $10^9$  variants feed machine-learning and deep-learning models that learn latent biophysical constraints. These models enable functional prediction and protein design, with iterative experimental validation closing the loop between prediction and observation.