

## Self-adaptive biosystems through tunable genetic parts and circuits

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## 1 **Abstract**

2 Biological systems often need to operate in complex environments where conditions can  
3 rapidly change. This is possible due to their inherent ability to sense changes and adapt their  
4 behavior in response. Here, we detail recent advances in the creation of synthetic genetic  
5 parts and circuits whose behaviors can be dynamically tuned through a variety of intra- and  
6 extra-cellular signals. We show how this capability lays the foundation for implementing control  
7 engineering schemes in living cells and allows for the creation of biological systems that are  
8 able to self-adapt, ensuring their functionality is maintained in the face of varying  
9 environmental and physiological conditions. We end by discussing some of the broader  
10 implications of this technology for the safe deployment of synthetic biology.

11

## 12 **Highlights**

- 13 • Tunable genetic parts allow for their input-output relationship to be dynamically varied in  
14 response to intra- and extra-cellular signals.
- 15 • Self-adaptive biological systems can be built using a combination of control engineering  
16 principles and tunable genetic parts and circuits.
- 17 • An ability to engineer self-adaptive systems will be crucial in deploying synthetic biology  
18 into complex and changeable real-world environments.

## 19 Introduction

20 A key characteristic of all living organisms is their ability to adapt. From altering metabolism  
21 to best utilize a shift in nutrients, to regulating ion transport to maintain cellular homeostasis,  
22 adaptive responses are crucial to many aspects of life. To enable such adaptive processes,  
23 cells have evolved a wide array of sensors able to capture information about their local  
24 environment as well as their internal state. These sensors are connected to cellular circuits  
25 that both monitor and modify internal processes with the goal of maintaining a desired  
26 functionality (e.g. homeostasis) no matter the perturbations experienced by the cell.

27 Unlike in Nature, engineered biological systems often lack the ability to adapt to  
28 changing conditions, making them fragile and causing them to break easily [1–6]. This stems  
29 historically from an absence of genetic parts that can be used to dynamically tune the  
30 response of a system and the additional burden of implementing control processes on top of  
31 a basic functioning system. This view is, however, beginning to change [7–9]. Recent  
32 developments in synthetic biology have led to a wide variety of biological parts able to  
33 precisely regulate the transcription [10–14] and translation [15–19] of genes in response to  
34 diverse intra- and extra-cellular signals [20]. Furthermore, the benefits of exploiting control  
35 engineering principles to create robust biosystems is also becoming recognized [7–9,21,22].  
36 This stems from a growing need in many applications for reliable and guaranteed  
37 functionalities no matter the strain of cell used, or the environment deployed to [23].

38 In this work, we discuss some of the recent advances towards engineering self-  
39 adaptive biological systems. We begin by providing an overview of the wide variety of parts  
40 now available for sensing and tuning cellular behaviors and show some of the ways these can  
41 be used to create adaptive genetic circuits. We then discuss recent steps towards using these  
42 circuits to implement closed-loop feedback control within living cells to create self-adaptive  
43 systems and end by outlining some of the future applications that such capabilities could  
44 support.

45

## 46 Tunable genetic parts

47 To develop an adaptive system, it is necessary to be able to dynamically alter/tune the input-  
48 output relationship of parts within the system. These ‘tunable’ components come in many  
49 different forms, however, conceptually have a common structure (**Figure 1a**). Each tunable  
50 element consists of an input and output, and a further tuner input that is able to alter the input-  
51 output relationship in a useful way [24]. Input, output and tuner signals can take many forms,  
52 from gene expression rate to protein phosphorylation state. However, one of the most  
53 commonly used is transcriptional activity [3,6,25]. This is captured by the RNA polymerase  
54 (RNAP) flux along DNA and can be directed to particular points by positioning

55 promoters that control the transcriptional initiation of RNAP. This makes it simpler to connect  
56 individual parts by making the output promoter of one the input promoter of another [10,26].

57 While there are many ways that the behavior of biological parts can be tuned, the most  
58 widespread and easiest to apply is through the control of gene expression. By incorporating  
59 additional regulatory parts to modify the rate of transcription and/or translation of an output  
60 gene it is possible to create a tunable expression system (TES) that can vary the amount of  
61 output protein produced for a given input transcriptional activity [24]. As gene expression  
62 underlies many core cellular behaviors this approach is a flexible means to control a variety  
63 of biological functionalities in a dynamic and tunable way.

64 The core structure of a TES comprises of promoters acting as signals for the input and  
65 tuner, a gene that is expressed as output, and internal regulators that allow the input and tuner  
66 promoters to dynamically alter output protein expression rate (**Figure 1a**). For the input and  
67 tuner promoters, a variety of sensors with transcriptional outputs now exist to sense  
68 environmental conditions such as chemical concentrations [27] and light [28,29], as well as  
69 internal cellular states (e.g. stress responses) [9] and population level features like cell density  
70 through quorum sensing [30]. Similarly, many output genes exist that enable control of cellular  
71 behaviors from modifying their metabolic state [31–33] to cell movement [34] and even cell-  
72 to-cell communications [30]. The final component in the TES is the internal regulator used to  
73 modulate how transcriptional activity of the input promoter is transformed into a protein  
74 expression rate. To make this relationship a function of the transcriptional activity of the tuner  
75 promoter, numerous types of transcriptional and translational regulators can be used (**Figure**  
76 **1b**). These include: 1. toehold switches (THSs) where translation rate is controlled through  
77 expression of a trigger RNA that is able to disrupt secondary structures around the RBS of the  
78 output gene [15,16,24]; 2. small transcription activating RNAs (STARS) which interact with  
79 transcriptional terminators that are placed in the 5' untranslated region (UTR) of a gene and  
80 regulate premature RNAP termination [11,12,35]; 3. small RNAs (sRNAs) that can be  
81 designed to bind the ribosome binding site for a gene of interest and suppress translation  
82 initiation [18]; 4.  $\sigma$ /anti- $\sigma$  pairs where the anti- $\sigma$  protein is expressed by the tuner promoter to  
83 reduce the expression rate of input promoters driven by the cognate  $\sigma$ -factor [36,37]; 5. split  
84 T7 RNAPs where the input and tuner promoters express different halves and the gene of  
85 interest is connected to the cognate promoter of the RNAP [38]; and 6. other programmable  
86 transcription factors like CRISPRi/a [14], transcription activator-like effector nucleases  
87 (TALENs) [39] and zinc fingers (ZFs) [40] that can be expressed by the tuner promoter and  
88 interfere or enhance transcription initiation or elongation from the input promoter.

89 Although it is more common for the input and tuner promoters to be different, recently  
90 it has been shown that by using identical promoters to control both regulatory inputs in unison,  
91 more stringent control of a protein expression can be achieved as well as sharp digital-like

92 transitions between OFF and ON states [35,41] (Greco et al. bioRxiv doi:  
93 10.1101/2020.07.04.187500).

94 It should also be noted that other approaches to tuning gene expression have been  
95 developed. For example, two-component systems where phosphorylation rates can be  
96 modified by the concentration of specific kinases [42] and CRISPRi systems where the  
97 strength of repression is controlled by base mismatches in the guide RNA (gRNA) [9].  
98 However, in most cases tuning of such systems requires the physical modification of the  
99 encoding DNA making it impossible to dynamically regulate behavior.

100

## 101 **Adaptive genetic circuitry**

102 To implement more complex functionalities, it is often necessary to connect together many  
103 genetic parts into a circuit [6]. In other engineering fields such as electronics, specifying the  
104 connections between components would generally be sufficient to create a working system.  
105 This is due to electronic components having standardized operating ranges to ensure  
106 compatibility and reliable functionalities no matter the context they are used in. For example,  
107 complementary metal-oxide-semiconductor (CMOS) electronic logic gates expect inputs of 0–  
108 1.5 V for an OFF state and 3.5–5 V for an ON state. In biology, such standardization is difficult  
109 due to the diversity of biochemical components used and challenges in engineering them to  
110 ensure a common level of response [26,43]. Therefore, rather than imposing constraints on  
111 biology that are near impossible to implement, it is instead necessary to work with the diversity  
112 present and ensure that components connected have inputs and outputs that are ‘matched’  
113 to guarantee signals propagate correctly [6,20,44]. Many of the advances in automated  
114 genetic circuit design have revolved around ensuring parts perform consistently when used in  
115 different ways (e.g. insulating their function from varying genetic context [45,46]) and  
116 automating the selection of combinations of parts within a circuit such that their inputs and  
117 outputs are optimally matched [6,26,43].

118 Tunable genetic parts can greatly simplify this process by removing the need to  
119 reassemble a circuit if two parts are found to be mismatched when connected. At the cost of  
120 additional tuning inputs to a circuit, tunable genetic parts can have their response function  
121 dynamically varied after circuit assembly (**Figure 2a**). This allows parts to be dynamically  
122 matched and opens up the possibility of rapidly optimizing overall circuit function without the  
123 need to reassemble underlying DNA (**Figure 2b**). In addition to simplifying the creation of  
124 optimized circuits, the ability to dynamically vary the response dynamics of individual parts is  
125 also valuable for systems that must function in highly changeable environments, where shifts  
126 might cause physiological changes that impact some or all parts in a circuit [4,5,47].

127           Beyond the tuning of steady-state response functions, circuits capable of exhibiting  
128 dynamic behaviors such as oscillations have also been developed, where characteristics such  
129 as period and amplitude can be varied through diverse inputs to the system. In one such  
130 oscillator for *Escherichia coli* cells, positive and negative feedback loops are implemented  
131 using the  $P_{BAD}$  (positive) and  $P_{tac}$  (negative) systems which can be further regulated using  
132 arabinose and isopropyl  $\beta$ -D-1-thiogalactopyranoside (IPTG), respectively [48]. It was found  
133 that increasing the concentration of Arabinose caused a lengthening of the oscillatory period,  
134 while increasing an IPTG concentration or temperature led to a shortening of the oscillatory  
135 period. Other tunable oscillator circuits have also been developed to allow for control via light  
136 [49], to synchronize behaviors across a population of cells [50] and to function in mammalian  
137 cells [51]. Furthermore, they've been modelled to demonstrate regulatory motifs capable of  
138 having their oscillatory amplitude and frequency tuned independently [52].

139

### 140 **Towards self-adaptive systems**

141 A limitation of using tunable genetic parts and circuits is the need for external inputs to be  
142 continually provided. A solution to this is to connect the output of a cellular process to the tuner  
143 input of the circuit, creating a closed-loop self-adaptive system. There has been growing  
144 interest in the application of closed-loop feedback control in biology and the role that control  
145 engineering principles might play in creating robust biosystems [8,21,22,53].

146           Some simple feedback control schemes have already been implemented in living cells.  
147 Many of these focus on the development of dynamic regulatory schemes for metabolism to  
148 maximize the yield of desired products [33,54,55]. Feedback is created by either using  
149 endogenous transcription factors that respond to intermediate metabolites of interest [31,56],  
150 or by the design of RNA aptamers able to sense and then actuate gene expression or shifts  
151 in metabolic fluxes in response to changes in metabolite concentrations (Glasscock et al.  
152 bioRxiv doi: 10.1101/529180). Related to this, general cellular stress responses have also  
153 been used as triggers for feedback control. Specifically, the  $\sigma^{32}$  heat-shock response of *E. coli*  
154 was found to be rapidly activated when cells were burdened by excessive protein expression  
155 [57]. By connecting the endogenous  $P_{htpG1}$   $\sigma^{32}$ -promoter to a CRISPRi based feedback control  
156 system (**Figure 3a**), it was shown that protein expression of burdensome synthetic genetic  
157 constructs could be dynamically regulated to reduce cellular burden [9]. This both increased  
158 overall protein yield, as there was less impact of cellular growth, and the evolutionary stability  
159 of the synthetic genetic constructs as there was less selective pressure for mutations. Similar  
160 approaches have been implemented using repressor proteins for negative feedback regulation  
161 and the  $P_{ibpAB}$   $\sigma^{32}$ -promoter as a sensor of burden [58]. Dynamic regulation of protein  
162 expression has also been performed in mammalian cells using translation-based negative

163 feedback control [59] and general-purpose gene expression controllers based on quorum-  
164 sensing [32].

165 More general feedback control schemes in living cells include the antithetic integral  
166 controller motif that uses sequestration mechanisms such as molecular titration to implement  
167 an embedded feedback controller [8]. This motif guarantees perfect adaptation, rejecting  
168 constant disturbances so that the output of the genetic system of interest initially responds to  
169 an external input but then returns to basal levels while the input persists [21].

170 Molecular titration has also been shown to be an effective mechanism to implement  
171 ‘comparator’ devices able to produce an output function of the mismatch between the levels  
172 of two different inputs, an essential component of any biomolecular controller [36,60].  
173 Implementations of more sophisticated control strategies have also been recently presented,  
174 such as the biomolecular PID controller presented in [61]. As the complexity of biomolecular  
175 control designs increases, to successfully implement the control function, the parts needed to  
176 construct the control strategy must be finely tuned to guarantee the right balance between the  
177 sensing and actuating parts of a circuit [62]. The use of tunable parts could open the way to  
178 the development of adaptive biomolecular controllers able to self-tune themselves in order to  
179 guarantee the robust execution of the control task they are assigned to perform even in the  
180 presence of perturbations, cell-to-cell variability, etc. This might be even more crucial when  
181 the control functions are spread among different populations in a microbial consortium as  
182 recently suggested in [63].

183 Beyond simple feedback motifs, it can be difficult to implement complex control  
184 algorithms using biochemical components because the feedback strengths and dynamics  
185 required may be difficult to match to available parts. Therefore, an intermediate step is  
186 sometimes taken whereby a computer is used to implement controller logic within the  
187 feedback loop and create what is termed a cybergenetic system [22] (**Figure 3b**).  
188 Cybergenetic systems often rely on single-cell microscopy platforms and microfluidics to image  
189 engineered cells whose current state is displayed via fluorescent reporter proteins and use  
190 chemical inducers [64,65] or light [28,29,66,67] as inputs to perturb the cells states in a pre-  
191 defined way (i.e. the cells are engineered to sense and update their state in response to a  
192 stimulus). The computer-based controller runs in real-time analyzing microscopy images to  
193 extract the current states of cells and then immediately computes a control action, which is  
194 then administered by varying chemicals concentrations or light that the cells are exposed to.  
195 Such systems have been shown capable of controlling both population [8,64,66,68] and  
196 single-cell behaviors [67]. Moreover, toolkits have emerged to simplify their creation by  
197 handling image analysis, tracking and calculation of control actions (Pedone et al. bioRxiv doi:  
198 10.1101/2020.06.25.171751). The major advantage of this hybrid approach is that the

199 computer controllers are cell-agnostic, allowing them to be used with any biosystem that has  
200 the same types of control inputs and observable outputs.

201

## 202 **Conclusions**

203 The creation of self-adaptive biosystems that can function in the face of varying and uncertain  
204 environments will be a crucial step for the safe deployment of synthetic biology into everyday  
205 life. Recent advances in biological control engineering provide the theoretical foundations  
206 necessary to design such systems and, as we have shown, tunable genetic parts and circuits  
207 can support their physical implementation [24]. While the self-adaptive systems built to date  
208 have mostly been small-scale proof-of concepts, it is clear that the ability to synthesize and  
209 assemble entire genomes is in reach [69,70]. Demonstrating the value of integrating tunable  
210 parts and circuits within these cellular systems will be crucial to moving beyond the mere  
211 recoding of existing genomic information and towards the creation of synthetic cells built from  
212 the ground up to reliably implement novel functionalities. Furthermore, they will support the  
213 scale-up of these systems by enabling us to move beyond single-cells and towards the  
214 engineering of collective behaviors of vast populations of cells [71] or even entire synthetic  
215 ecologies [72].

216

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223

## 224 **Author Contributions**

225 All authors contributed to the writing and editing. T.E.G. and V.B. produced the figures.

226

## 227 **Declaration of Interest**

228 None.

229 **References**

230 • of special interest

231 •• of outstanding interest

232

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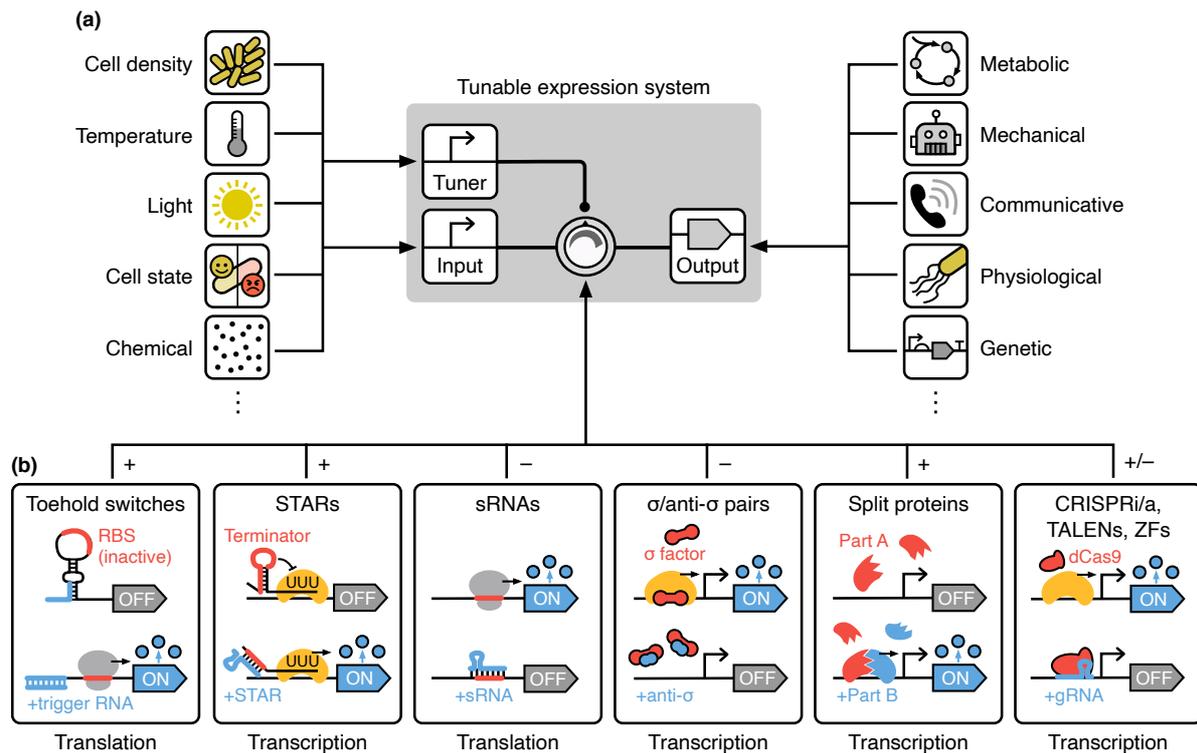
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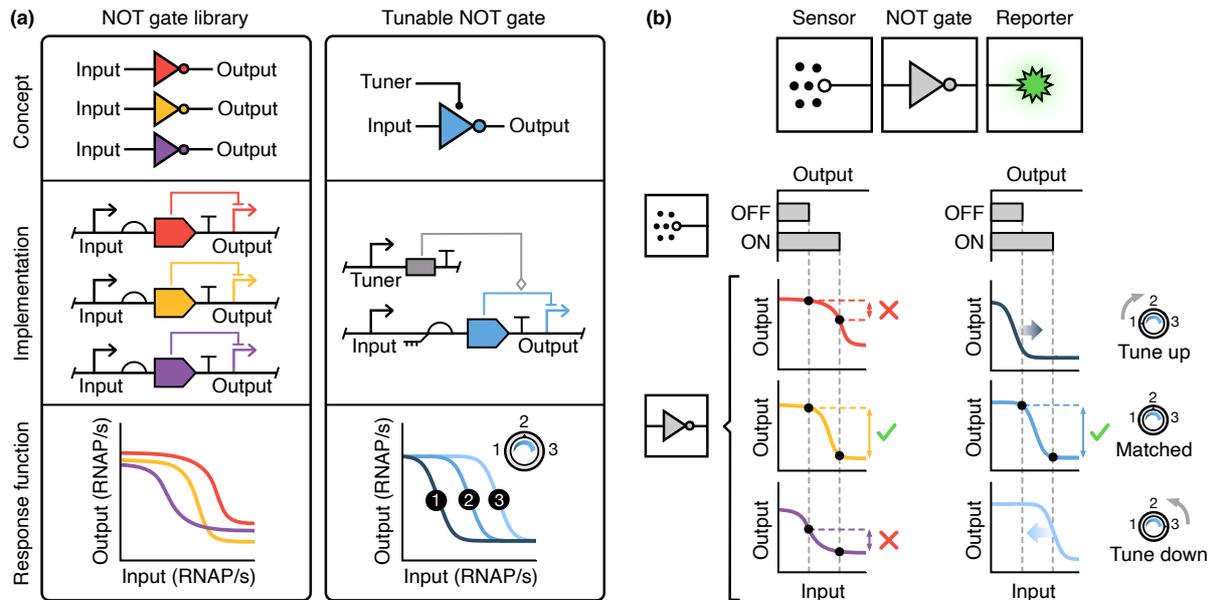
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470 **Figures and captions**

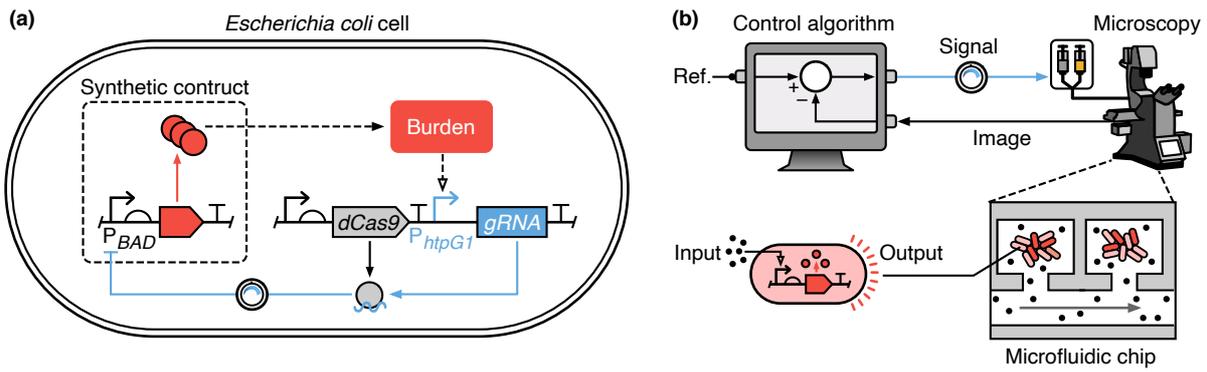
471

472 **Figure 1: Tunable genetic parts.** (a) Schematic of a tunable expression system (TES) where  
 473 a variety of different inputs and output can be selected. Typically, inputs are transcriptional  
 474 signals related to environmental or cellular states and the output is the expression of a gene  
 475 that influences cellular behavior or acts as an input to another part of a larger circuit. (b) Major  
 476 regulatory mechanisms that can be used to tune gene expression in a TES. Both active and  
 477 inactive states shown in addition to whether the tuner will cause activation (+) or repression  
 478 (-) of the output. The stage in gene expression (i.e. transcription or translation) where  
 479 regulation takes place is shown below the box for each mechanism. Ribosomes and RNA  
 480 polymerase (RNAP) shown by light grey and orange shapes without an outline, respectively.  
 481 For the CRISPRi/a, TALENs and ZFs box a repressive CRISPRi system is shown. This can  
 482 be modified to be an activator by fusing dCas9 to an activator domain to recruit RNAP to the  
 483 promoter. In general, the additional blue element would be expressed by the tuner input to  
 484 modulate expression of the output. RBS, ribosome binding site; sRNA, small RNA; STAR,  
 485 small transcription activating RNAs; siRNA, small interfering RNA; CRISPRi/a, clustered  
 486 regularly interspaced short palindromic repeats interference/activation; TALEN, transcription  
 487 activator-like effector nuclease; ZF, zinc finger; gRNA, guide RNA.



488

489 **Figure 2: Tunable genetic parts enable the construction of adaptive circuits.** (a) Libraries  
 490 of genetic parts (e.g. NOT gates) are commonly created that cover a range of different  
 491 behaviors (left box). These differences are shown by the specific response function of each  
 492 part, which captures the steady-state input-output relationship. For most genetic parts the  
 493 response function is fixed, and so physical replacement is necessary if a part is not compatible  
 494 when used in a system. In contrast, tunable genetic parts (right box) have additional tuner  
 495 inputs that allow the shape and position of the response function to be dynamically varied as  
 496 required. (b) Schematic of a simple genetic circuit where a sensor input is inverted to give a  
 497 desired output reporter (e.g. green fluorescence). For the sensor and NOT gate to parts to work  
 498 effectively, the output of the sensor must 'match' the response function of the NOT gate (dotted  
 499 grey lines). If the parts are matching, then a large change in the NOT gate output will occur  
 500 when the sensor switches between OFF and ON states. For standard NOT gates (left column)  
 501 entire libraries need to be assembled and screened to find a working combination.  
 502 Furthermore, if the environment changes then so too might the behavior of the parts making  
 503 reassembly necessary. For a tunable NOT gate (right column), the tuner input can be varied  
 504 until the gate perfectly matches the sensor's outputs. No reassembly is required, allowing the  
 505 circuit to be dynamically tuned to changing conditions. Genetic circuits shown using Synthetic  
 506 Biology Open Language (SBOL) Visual notation [73]. RNAP, RNA polymerase.



507

508 **Figure 3: Self-adaptive systems.** (a) Embedding burden-based controller. A synthetic  
 509 construct expresses a burdensome protein. Endogenous cellular processes (dashed arrows)  
 510 lead to the activation of the  $P_{htpG1}$  promoter under high levels of protein expression burden  
 511 causing expression of a guide RNA (gRNA). This gRNA forms a complex with a constitutively  
 512 expressed dCas9 protein that then targets the promoter of the synthetic construct, down  
 513 regulating its expression. The strength of this negative feedback loop is dynamically 'tuned'  
 514 by the endogenous burden signal as well as mismatches in the gRNA to the target promoter  
 515 that reduce the binding affinity of the dCas9:gRNA complex. Panel adapted from [9]. (b)  
 516 Schematic of an external *in silico* control system. Living cells grow in a microfluidic chip that  
 517 is continually imaged by a microscope. These images are set to a computer, analyzed and an  
 518 output signal from the cells (e.g. fluorescence) compared to a desired reference value. A  
 519 control algorithm assesses this difference and emits a control signal, which actuates syringes  
 520 and changes the concentration of a signaling molecule provided to the cells. The cells sense  
 521 this change and alter their gene expression in response. The strength of feedback in this  
 522 system can be tuned by modulating the control signals produced. Grey arrow in the  
 523 microfluidic chip represents the flow of nutrients and signaling molecules. gRNA, guide RNA.