

REVIEW

**Biology by design: reduction and synthesis
of cellular components and behaviour****Philippe Marguet¹, Frederick Balagadde², Cheemeng Tan³
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Biological research is experiencing an increasing focus predominantly on the application of knowledge rather than on its generation. Thanks to the increased understanding of cellular systems and technological advances, biologists are more frequently asking not only ‘how can I understand the structure and behaviour of this biological system?’, but also ‘how can I apply that knowledge to generate novel functions in different biological systems or in other contexts?’ Active pursuit of the latter has nurtured the emergence of synthetic biology. Here, we discuss the motivation behind, and foundational technologies enabling, the development of this nascent field. We examine some early successes and applications while highlighting the challenges involved. Finally, we consider future directions and mention non-scientific considerations that can influence the field’s growth.

Keywords: synthetic biology; gene circuits; biological design; genetic engineering;
computational biology; metabolic engineering

1. INTRODUCTION

Imagine you have been charged with building a robot capable of complex and autonomous operations in a dynamic environment. What are the most advantageous characteristics to build into such a machine? To perform work, energy will be needed—renewable energy extracted from the environment is ideal. To respond with meaningful behaviour, information gathering and possessing capabilities will be required. For coordinated operations, communication with other robots is essential. To maintain a long-term function, a self-contained repair or reproduction system will be necessary. To attempt some goals, the robot will need to be a minuscule. To achieve economic feasibility, production costs will have to be low. While all these requirements are significant hurdles to the robotics engineer on a budget, they are feats that life has accomplished time and time again.

Consider one of the simplest forms of life, bacteria. Only a few micrometres long bacteria are capable of many of the above requirements, including, entering minuscule environments, surviving on local nutrients and responding to fluctuations in their environment

with adaptive behaviour (such as chemotaxis (Falke *et al.* 1997), altered nutrient utilization (Jacob & Monod 1961) and temperature-dependent gene expression (Yura & Nakahigashi 1999)). Many bacterial species communicate in order to produce coordinated behaviour (Bassler & Losick 2006) and with doubling times as fast as 20 min, their reproduction capacity is remarkable.

In fact, an engineer building a device on a bacterial ‘chassis’ would only need to build one functioning prototype, culture overnight in low-cost media and return the next morning to obtain trillions of virtually identical copies. In a sense, this is like programming a minuscule but complex computer that can also reproduce. As appealing as this concept may seem, several fundamental questions arise: what functions are we capable of programming into a living organism? To what extent will these functions be performed predictably and robustly? What is the best way to implement a pre-defined design goal and what challenges and opportunities may arise? These are some of the questions that the burgeoning field of synthetic biology is beginning to address.

Over the past few years, synthetic biologists have generated remarkable systems including: an expanded genetic code in *Escherichia coli* (Wang *et al.* 2001); various logic gates (Dueber *et al.* 2003; Rackham &

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Chin 2005a); rewired yeast mating and osmolarity response circuitry (Park *et al.* 2003); bistable switches in bacteria (Gardner *et al.* 2000; Isaacs *et al.* 2003); yeast (Becskei *et al.* 2001) and mammalian cells (Kramer & Fussenegger 2005); photographic bacteria (Levskaya *et al.* 2005); genetic and metabolic oscillators (Elowitz & Leibler 2000; Atkinson *et al.* 2003; Fung *et al.* 2005); artificial communication in bacteria (Bulter *et al.* 2004) and yeast (Chen & Weiss 2005); and many other interesting and useful systems.

Although there is a debate about the scope and boundaries of the field, some advocates supply that ‘synthetic biology’ is:

(A) the design and construction of new biological parts, devices and systems and (B) the re-design of existing, natural biological systems for useful purposes.

(www.syntheticbiology.org,
syntheticbiology.org/FAQ).

It is worth examining this definition more closely. Inherent in part (A) are engineering principles—the notions of abstraction and hierarchy. One level of abstraction consists of biological components with simple albeit well-defined functions, operating under defined conditions, i.e. *parts*. At a higher level of abstraction, parts can be combined to form *devices*. Similarly, devices come together to form *systems* on a third level of abstraction. The basic premise is that an individual researcher can work at one of these levels without necessarily requiring to know the precise mechanics of operation at another level (Endy 2005).

Part (B) states that biology is being redesigned for ‘useful purposes’. What purposes you might wonder? The first purpose may be obvious, and it is the practical application of biologically modified organisms in human life. Although our ancestors did not possess the advanced genetic tools available today, the litany of domesticated species including fermentation yeasts, crop grains and silkworms is a testament to the vast utility of modified living organisms to humans. However, modification of living organisms by traditional means, i.e. artificial selection, is an incremental and slow process with limited pay-offs during an individual’s lifetime. For example, it has taken approximately 15 000 years of domestication by selective breeding to turn wolves into present-day dogs (Leonard *et al.* 2002), a process which grouped desirable genes in particular breeds. Improvements in DNA synthesis and genomic engineering methods have enabled the introduction of genetic changes in relatively short time frames. Such technologies will engender the practical application of modified biological systems to new areas, such as therapeutics, renewable energies and others. The practical applications of modified biological systems represent the first useful purpose behind a redesign.

Of course, even possessing large-scale DNA technology capable of making the changes needed to produce a guide dog from a wolf is not enough. The necessary DNA changes have to be known in advance in order to be made. This is far from the case—especially for a complex organism like the dog. Comparative genomics can elucidate the differences between the

organisms, but does not yield the full understanding needed to prospectively say ‘If I want to program guide animal functions into organism X, here are the changes I will make and this is how those changes work.’ In the venerable words of physicist Richard Feynman, ‘what I cannot create, I do not understand’ (Hawking 2001). The laws of physics and chemistry apply to living systems just as they apply to non-living things, such as mechanical engines. Yet, designing and constructing even simple biological systems remain a major challenge, whereas mechanical engines can be predictably engineered. Feynman would conclude that there must exist fundamental gaps in our understanding of how biological systems operate. Synthetic biology is exploring these gaps in understanding by attempting to build and apply such systems.

Scientific experiments are run under specific conditions with the hope that the conclusions drawn will be applicable in a broader context. The creation of biological systems by using currently accepted (or debated) principles would test the limitations and applicability of those principles. Likewise, implementation of existing genes, proteins and pathways in non-native settings can help elucidate their functions and reveal unknown requirements for their operation. Synthetic biologists therefore aim not only to produce interesting and useful designs, but also to simultaneously develop a greater understanding of biological components and design principles in general (Sprinzak & Elowitz 2005). Therefore, the second, and equally important, purpose of synthetic biology is to gain the biological insight that arises from testing our knowledge during the design and implementation process.

2. FOUNDATIONAL TECHNOLOGIES

Just as the development of the microscope made the discovery of cells possible (Dunn & Jones 2004), new technologies are providing the critical foundation needed for synthetic biology. Here, we discuss the following four major advances that have produced enabling tools for experimentation and analysis in this regard: DNA synthesis; parts and devices design and optimization; systems modelling; and observational capabilities. For an overview of where these technologies interact with synthetic biology (figure 1).

2.1. DNA synthesis

At the core of every living thing, dictating that organism’s characteristics and behaviour is a string of nucleotide bases—its DNA. To reprogram an organism, that DNA needs to be altered or supplemented. Until recently, DNA manipulations were almost exclusively done in a ‘copy, cut and paste’ manner using polymerases, restriction endonucleases and ligases, respectively. While this enzymatic approach has produced a wealth of scientific advances, implementing a complicated biological design by these means is the literary equivalent of writing a paper using a photocopier, scissors and a stick of glue. Recently, however, biologists have received their metaphorical ‘typewriter’. Tian and colleagues developed a large-scale

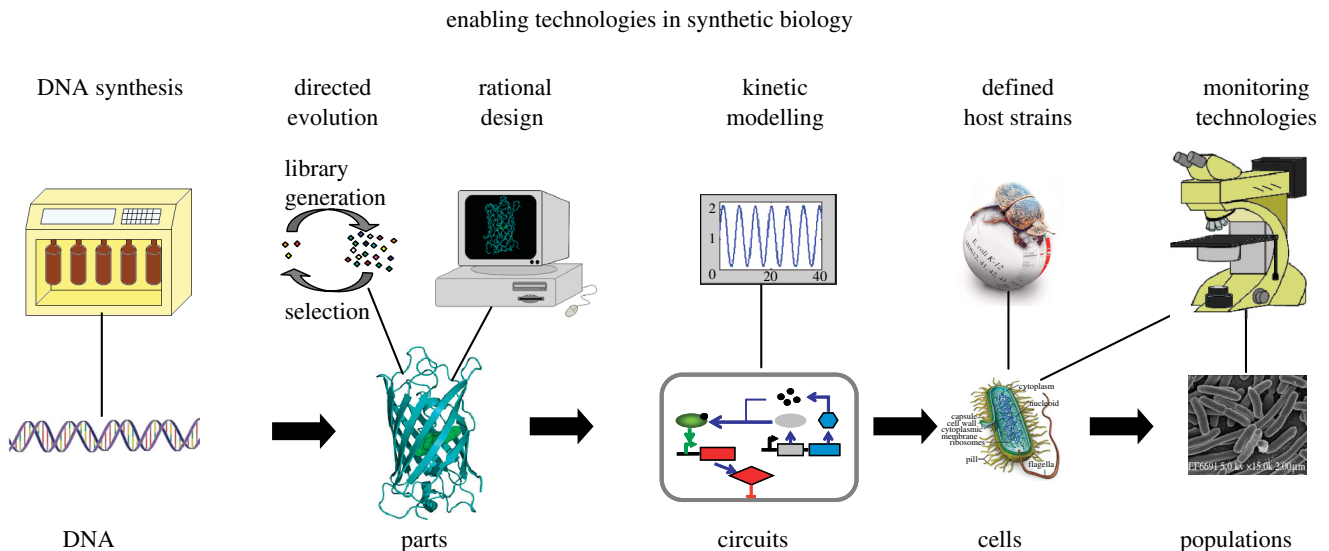


Figure 1. Interplay between engineering tools with a biological hierarchy. In order to simplify biological design, it is valuable to use an abstraction hierarchy. DNA codes for ‘parts’ that interact with each other to form circuits. The totality of all circuits and structures forms a single cell, which interacts with its neighbours and environment to form a population. At each level, technologies have been developed to assist and enable design. Shown here are the major advances that significantly reduce longstanding design, analysis and production barriers. Together, these technologies are helping to make integrated biological design a reality.

DNA synthesis method by using parallel oligonucleotide synthesis on a programmable microfluidics chip, followed by PCR amplification (Tian *et al.* 2004). In order to reduce the error rate to 1 in 1394 bases, the authors hybridized their ‘construction oligos’ against complimentary ‘selection oligos’ and washed away mismatches. Construction oligos were assembled into larger genes using polymerase assembly multiplexing (PAM), an overlap PCR-based method. Using their chip-based technology, the authors simultaneously synthesized and optimized all 21 genes encoding the 30S ribosomal subunit from *E. coli*.

In a different study, Jacobson and co-workers developed a method that can further improve the removal of error-containing DNA fragments (Carr *et al.* 2004). Using this method, which exploits the gel mobility shift apparent when MutS binds a mismatched double-stranded DNA, they were able to obtain an error rate as low as 1 in 10 000 (the average length of a prokaryotic gene is 924 bp, while that of a eukaryotic gene is 1346 bp; Xu *et al.* 2006). By applying these and other technologies, commercial companies are now able to offer large-scale (multi-kilobase and up) DNA synthesis for under \$1, a base with a two- to four-week turnaround time. Whereas these prices do not make a 10 kb construct inexpensive for most researchers, they imply that commercial synthesis has begun to rival the equipment, materials, labour and validation costs incurred by traditional cloning and construction means for select applications. It may be possible to further reduce these costs by 10–100-fold, i.e. 1–10 cents per base, within the next decade by fully automating and streamlining new high-throughput techniques (J. Tian, personal communication).

Total DNA synthesis can be used to alter or improve the sequences being built. In traditional cloning, targeted mutational changes are made only to small

regions (approx. 20 bases) at once. Furthermore, each region altered imposes additional experimental steps. DNA synthesis methods, however, can synthesize an altered sequence with no more effort than that necessary to synthesize a wild-type sequence of the same length. For example, a protein-coding sequence can be matched with regard to codon usage in the host organism where it will be expressed. In this case, the sequence of amino acids in a protein is left unaltered by the modification, but translation efficiency can be improved by using codons whose cognate tRNAs are more abundant. Similarly, the sequence can be altered to remove or create mRNA secondary structures without changing the resulting amino acid sequence. Furthermore, a gene whose sequence is known, but whose DNA is hard to obtain, can be easily synthesized.

2.2. Design and optimization of parts

One level of abstraction from the DNA synthesis and manipulation is the parts production, which can be accomplished through either rational design or directed evolution. Recently, improved algorithms and processor power have allowed computational design efforts to achieve new milestones in reprogramming the function of many well-characterized natural proteins. In a series of studies integrating both computation and experiments, the Hellinga laboratory succeeded in introducing an allosteric control switch into the proton-ATP pump (Liu *et al.* 2002); retooling sugar-sensing receptors to bind novel ligands, such as lactate, trinitrotoluene (TNT) and serotonin (Looger *et al.* 2003), and converting a receptor into a functional triose phosphate isomerase enzyme (TIM), catalysing a 10^5 – 10^6 -fold rate improvement over the uncatalysed reaction (Dwyer *et al.* 2004). Significantly, they even demonstrate that designed parts are active *in vivo* and

379 can be used to produce more complex systems. The
 380 TNT receptor and a designed Zn receptor were shown
 381 to induce gene expression in response to exogenous
 382 ligands when implemented in some of the earlier
 383 reported examples of synthetic signalling pathways
 384 (Dwyer *et al.* 2003; Looger *et al.* 2003). Likewise, the
 385 TIM enzyme was sufficiently active to complement its
 386 wild-type version and restore viability under gluconeo-
 387 genic conditions.

388 Computational design has also found applications
 389 beyond altering the specificity or the enzymatic
 390 function. For example, the Baker laboratory has
 391 designed a new protein that folds to form a novel
 392 structure—matching their modelling predictions
 393 (Kuhlman *et al.* 2003). They also apply their compu-
 394 tational methods to increase the thermostability of an
 395 enzyme by identifying key mutations. When some
 396 mutations were applied in concert, the result was a
 397 30-fold increase in half-life at 50°C (Korkegian *et al.*
 398 2005). The examples here illustrate altered specificity,
 399 novel functions and structures, improved stability and
 400 introduction of allosteric control. They highlight some
 401 of the contributions that computational protein design
 402 has made for parts generation and improvement. While
 403 we have only drawn examples from two research
 404 groups, computational protein design is a vast and
 405 growing field, with important contributions made by
 406 numerous other laboratories (Park *et al.* 2004).

407 Applying rational design to parts alteration or
 408 creation is advantageous, in that it cannot only
 409 generate products with defined function, but it can
 410 also produce biological insights into how the designed
 411 function comes about. However, it requires prior
 412 structural knowledge of the part, which is frequently
 413 unavailable. Directed evolution is an alternative
 414 method that can effectively address this limitation by
 415 allowing parts engineering without design. In essence,
 416 directed evolution begins with the generation of a
 417 library containing many different DNA molecules, often
 418 by error-prone DNA replication, DNA shuffling or
 419 combinatorial synthesis. The library is then subjected
 420 to high-throughput screening or selection methods that
 421 maintain a link between genotype and phenotype in
 422 order to enrich the molecules that produce the desired
 423 function. The process is then iterated to approach a
 424 desired endpoint (Arnold 2001; Kolkman & Stemmer
 425 2001; Joyce 2004). A recent example of parts creation
 426 by directed evolution is the expansion and alteration of
 427 LuxR specificity for acyl-homoserine lactone ligands
 428 (Collins *et al.* 2005, 2006). LuxR is a transcriptional
 429 activator from the marine bacteria *Vibrio fischeri*, and
 430 it is naturally responsive to the signalling molecule
 431 3OC6HSL. Collins *et al.* first employed a screening
 432 scheme to identify mutations that broadened the
 433 binding specificity of LuxR to other small molecules
 434 in the same class as 3OC6HSL (Collins *et al.* 2005).
 435 They then used a dual-selection method (Yokobayashi &
 436 Arnold 2005) to redirect LuxR specificity to one of those
 437 molecules, C10HSL (Collins *et al.* 2006). The result was
 438 a new protein that responds to the second chemical, but
 439 no longer to the first. These parts may be particularly
 440 beneficial to designers desiring multiple channels of
 441 simultaneous communication between cells. Directed

evolution can also be applied at other levels of biological
 hierarchy, for example, to evolve entire gene circuits
 (Yokobayashi *et al.* 2002).

Rational design and directed evolution should not be
 viewed as opposing methods, but as alternate ways to
 produce and optimize parts, each with their own unique
 strengths and weaknesses. Directed evolution requires a
 high-throughput way to both screen and select for a
 desired function and that functional mutants exist in
 the sequence space sampled. This second constraint
 becomes less likely as the desired function diverges
 further from the initial function. On the other hand,
 while rational design strategies can make multiple
 changes or large-scale alterations that incorporate
 scientific knowledge, these strategies are rarely precise
 enough to finely tune the system behaviour. Further-
 more, it is difficult to know if additional optimization is
 possible when employing rational design. For these
 reasons, both methods can and should be used in
 conjunction and will hopefully continue to be applied in
 unison during the years to come.

Recent years have witnessed increasing interest in
 using parts based on RNA for intricate control of
 gene expression (Davidson & Ellington 2005; Isaacs
et al. 2006). One particular line of research has been
 largely inspired from metabolite-controlled ribos-
 witches prevalent in nature (Mandal & Breaker
 2004; Nudler & Mironov 2004). RNA switches are
 advantageous in their fast response, broad applica-
 bility and chemical nature. RNA switches contain a
 ligand-binding region, or aptamer domain, that
 controls the function of an effector domain through
 binding-induced conformational changes. Strategies
 for the evolution of RNA aptamers and functional
 RNAs were developed early on (Ellington & Szostak
 1990; Robertson & Joyce 1990; Tuerk & Gold 1990)
 due to the fact that the same molecule plays both
 functional and information-encoding roles (i.e. the
 genotype–phenotype link required for directed
 evolution schemes is intrinsic to the molecule). This
 allows the generation of a library directly from the
 products of a competitive screen in the previous
 round. Furthermore, the entire selection, amplifi-
 cation and iteration procedure can be economically
 accomplished *in vitro*. The chemical nature of RNA,
 with four bases possible at each position, means that
 a higher percentage of available sample space can be
 covered while evolving an RNA molecule than a
 protein of similar length (20 amino acid possibilities
 per position). Additionally, the interactions within an
 RNA molecule are largely driven by complementary
 base pairing. As a result, relatively accurate methods
 for the secondary structure prediction of RNA have
 been developed and are widely used (Mathews *et al.*
 1999; Zuker 2003). Secondary structure information
 is valuable, because it can allow a researcher to make
 rationally guided changes.

In a recent work from the Smolke group (Bayer &
 Smolke 2005), switches were developed that exposed an
 antisense stem sequence upon binding a ligand, produ-
 cing a riboregulator. Ligands, such as theophylline,
 controlled switches that turned on gene expression, as
 well as switches that turned off gene expression. These

switches were shown to be tunable by making simple changes to the RNA sequence guided by thermodynamic properties. Multiple switches functioned independently in yeast even when binding similar molecules. Switches such as these may be useful in sensing cellular conditions and could also act as feedback mechanisms for tuning metabolic pathways in response to the depletion or accumulation of reactants, intermediates or products. Gallivan and colleagues demonstrate a synthetic RNA switch that is functional in prokaryotes and can be applied in screening or selection schemes that tie *in vivo* levels of small molecules to a reporter gene or cell survival, respectively (Desai & Gallivan 2004). In this manner, one could screen enzyme libraries for a desired catalytic function. Inversely, if the small chemical is supplied, then a library of riboswitches could be screened for binders that alter gene expression. Suess and colleagues, who first described a rationally designed *in vivo* RNA switch, implement it in such a way that it functions as a logic gate with another ligand, xylose (Suess *et al.* 2004). Perhaps, the best-known form of gene regulation by RNA, however, is the role of interfering RNA (Hannon 2002). Yokoboyashi and colleagues show that it is possible to modulate shRNA activity through the action of a small chemical by fusing the shRNA to an aptamer responding to the chemical (An *et al.* 2006).

Synthetic riboregulators need not be ligand controlled. Collins *et al.* demonstrate a general method to introduce RNA-mediated post-transcriptional regulation into prokaryotic genes (Isaacs *et al.* 2004). They introduced a short sequence between the promoter and ribosome-binding site that when translated into mRNA folds into a hairpin with the adjacent ribosome-binding site, sequestering the site and preventing translation. Translation can be restored by expressing a trans-acting RNA that binds the hairpin and forms a more stable structure, which frees the ribosome-binding site.

These examples demonstrate that the cellular engineer of the future will not be restricted only to simply combine the catalogue of known biological parts, but will also have the tools needed to supplement natural parts with custom made parts for specific applications.

2.3. Modelling-guided circuit engineering

The engineering process usually involves multiple cycles of design, optimization and revision (box 1 and figure 2). This is particularly apparent in the process of constructing gene circuits. As the number of interacting parts and reactions increases, it becomes more difficult to intuitively predict circuit behaviour. Towards these ends, mathematical modelling is a useful design tool, in particular, for systems with complex dynamics, such as bistability and oscillations. The importance of mathematical modelling has been increasingly appreciated, as evidenced by its extensive application in systems biology as a way to decipher ‘design principles’ of natural biological systems (Asthagiri & Lauffenburger 2000; Tyson *et al.* 2001; Gilman & Arkin 2002; You

2004). In comparison, the utility of modelling in synthetic biology seems even more dominant (Hasty *et al.* 2002; Kaern *et al.* 2003).

Various mathematical formulations can be used to model gene circuits. At the population level, gene circuits can be modelled using ordinary differential equations (ODEs). In an ODE formulation, the dynamics of the interactions within the circuit are deterministic. That is, given the same initial condition and numerical configurations, different rounds of simulations will lead to exactly the same results. In other words, the ODE formulation ignores the randomness intrinsic to cellular processes and is convenient for circuit designs that are thought to be less affected by noise or when the impact of noise is irrelevant. For instance, ODE models have been used to guide experimental efforts to program population dynamics in the temporal domain (You *et al.* 2004; Balagadde *et al.* 2005) or the spatial domain (Basu *et al.* 2004, 2005). Importantly, an ODE model facilitates further sophisticated analyses, such as sensitivity analysis and bifurcation analysis. Such analyses are useful to determine how quantitative or qualitative circuit behaviours will be impacted by changes in circuit parameters; this has been almost a standard practice in engineering of most gene circuits accomplished so far (box 1). For instance, in designing a bistable toggle switch, bifurcation analysis was used to explore how qualitative features of the circuit may depend on reaction parameters (Gardner *et al.* 2000). Results of the analysis were used to guide choice of genetic components (genes, promoters and ribosome-binding sites) and growth conditions to favour a successful implementation of designed circuit function.

In a single cell, however, a gene circuit’s dynamics often involve small numbers of interacting molecules. Such small numbers will result in highly noisy dynamics even for expression of a single gene (Elowitz *et al.* 2002; Ozbudak *et al.* 2002). For many gene circuits, the impact of such cellular noise may be critical and needs to be considered. This can be done using stochastic models (Rao *et al.* 2002). Different rounds of simulation using a stochastic model will lead to different results each time, which presumably reflect aspects of noisy dynamics inside a cell. For synthetic biological applications, the key of such analysis is not necessarily to accurately predict the exact noise level at each time point. This is not possible even for the simplest circuits due to the ‘extrinsic’ noise component for each circuit (Elowitz *et al.* 2002). Rather, it is a way to determine to what extent the designed function can be maintained and, given a certain level of uncertainty or randomness, to what extent additional layers of control can minimize or exploit such variations. For instance, a number of computational studies have been conducted to analyse the potential of cell–cell communication to synchronize intrinsically noisy and unreliable oscillators (McMillen *et al.* 2002; Garcia-Ojalvo *et al.* 2004).

Mathematical models, either stochastic or deterministic, can be digitally ‘evolved’ *in silico* to generate optimal circuit designs that satisfy a particular objective. Francois and Hakim used genetic algorithms

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Box 1.

A recipe for engineering gene circuits (also see figure 2)

Design

(i) Determine the design goal

For the purposes of this tutorial we will attempt to construct a population of cells that restricts its cell density below that imposed by nutrient limitations (You et al. 2004). The implementation discussed below is a revised version (Balagadde et al. 2005).

(ii) Pick suitable host organisms/strains

Key characteristics to consider here are: ease of genetic manipulation, growth rate, survivability under the desired conditions, and endogenous machinery you wish to exploit. *E. coli* could be used for this application.

(iii) Identify necessary ‘parts’

Available places to draw from include the literature, genome sequences, colleagues, and the MIT registry (The_BioBricks_Foundation). Recall that: (i) Parts need not be from the host organism. While native parts are likely to function properly, they can lead to crosstalk with endogenous systems. (ii) Parts need not exist; they can be developed by rational design or directed evolution. (iii) The better characterized the parts, the easier your job will be. (iv) It is advantageous to include parts as reporters. In this tutorial we will pick the quorum sensing genes *luxR* and *luxI*, as well as the toxin gene (CcdB) from F plasmid segregation.

Modeling

(iv) Build a mathematical model

Start with the simplest model that can capture the circuit dynamics (for example a simplifying assumption might be to assume a protein’s production rate depends on a transcription factor rather than explicitly modeling mRNA production, translation, and decay).

(v) Explore circuit dynamics *in silico*

Address questions like: can the network architecture give you the function you want? What parameters are most critical for success? How do circuit dynamics change with parameters?

Implementation, testing, and debugging

(vi) Determine the DNA implementation of your circuit.

In our case we will implement our circuit on a plasmid and need decide on copy number, what promoters, RBSs, transcription terminators, and perhaps degen tags to use. Another choice at this time is to decide if any components need to be expressed together on a polycistronic RNA. In this example, the circuit is implemented in a medium copy number plasmid (p15a origin) which the *luxR* and *luxI* gene are co-expressed by a P_{lac/ara} promoter. The CcdB gene is controlled by a P_{luxI} promoter. Kanamycin resistance is used as a selection marker.

(vii) If modeling indicates that a particular parameter is critical, build multiple versions

It is rare for all parameters to be perfectly balanced on the first experimental implementation. Designing multiple circuits at once to sample a critical parameter space can increase the chance for initial success. It may also yield interesting information about whether that particular parameter is truly critical.

(viii) Test your circuit and decide whether to retest, revise, or redesign

If it works as predicted you can continue to fully characterize it. If not, can you fit your model to explain the behavior that is observed? What parameters may need altering to generate the desired function? At this point you can: (i) redesign the circuit to address critical parameter changes, and perhaps ‘fine tune’ the circuit function by directed evolution; or (ii) test the circuit in other strains or growth conditions.

A working design usually requires multiple rounds of iteration of steps listed above, which is often the most time consuming portion of biological design.

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to design gene regulatory networks that exhibited hysteresis or oscillations (Francois & Hakim 2004). Initially, a pool of gene circuits was constructed from basic reactions representing activation, repression and post-translational modification. These circuits were subsequently evolved using numerical simulations to obtain a desired output by repeated rounds of digital ‘mutations’ and functional ‘screening’. Several unique designs were generated that satisfy each design goal. These designs could serve as alternatives to consider, model or test during the circuit engineering cycle.

One of the most exciting aspects of synthetic biology is the multiple avenues being used to address questions. While some researchers may only apply a particular method for a given application, the domain as a whole

will benefit from the use of these complementary approaches. For example, a simple linear cascade can be implemented using transcriptional regulation or reversible protein modification, both of which are prevalent in nature. Implementation by transcriptional control is appealing, because it is generally easier to stitch multiple DNA elements together. However, multi-component transcriptional cascades can introduce a significant time delay, as shown by Hooshangi et al. (2005). In this work, a one-stage cascade reached its half-maximal activation in minutes, whereas a three-stage cascade took several hours. Rosenfeld & Alon found that long transcriptional cascades are rare in the sensory systems of relatively short lived *E. coli* and *Saccharomyces cerevisiae* (Rosenfeld & Alon 2003).

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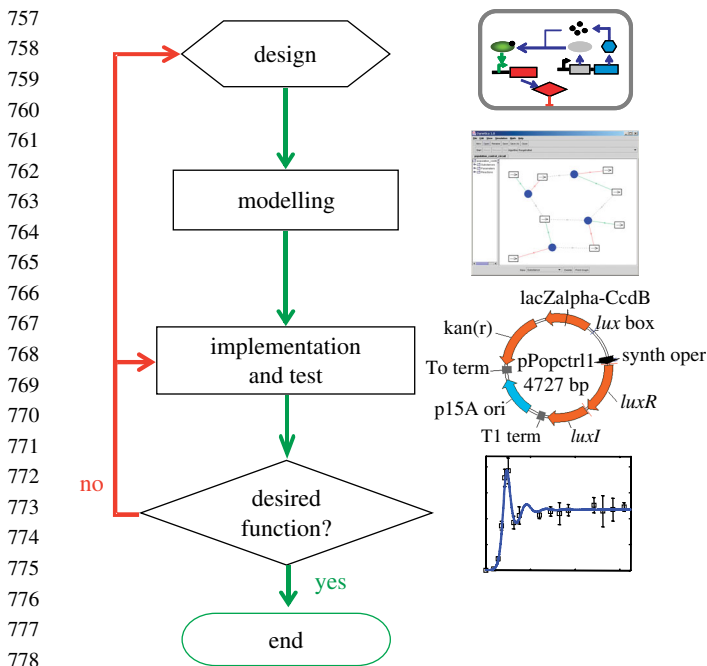


Figure 2. The typical process for engineering gene circuits (see box 1 for more details).

Protein modification-based circuits can offer much faster temporal response (Kholodenko 2006). As the field matures, it is probable that synthetic circuits, like nature, will integrate both DNA and protein regulatory logic in their design. The combination will exploit advantages of each method while mitigating their weaknesses. These choices will require mathematical modelling to ensure that the circuits can perform on the desired time-scale for a particular operation.

In most attempts to engineer gene circuits, mathematical models are often purposefully simplified to capture the qualitative behaviour of the underlying systems. Simplification is beneficial partially due to the limited quantitative characterization of circuit elements, one limitation that the BioBricks project aims to address (The BioBricks Foundation, Registry of Standard Biological Parts.), and partially because simpler models may better reveal key design constraints. The caveat, however, is that a simplified model may fail to capture richer dynamics intrinsic to a circuit. When engineering a population controller, we built a highly simplified kinetic model to capture the essence of the circuit dynamics, including cell growth, signal accumulation, killer protein accumulation and subsequent cell killing. The model predicts that the system will always lead to a stable regulated state, and this prediction was supported by the observations made in batch cultures (You *et al.* 2004). Yet, later, we observed sustained oscillations when cells expressing the circuit were grown in a microchemostat (Balagadde *et al.* 2005). One way to reconcile the experimental and modelling results was to introduce an extra step of regulation in our model, which indeed resulted in sustained oscillations for biologically feasible parameters. We note that still more layers of regulation are involved, further complicating the modelling analysis (figure 3).

2.4. Culturing and monitoring technologies

To determine if a synthetic circuit works as designed, one must be able to test it and observe its dynamics. These tasks have benefited from the rapid development of improved culturing and observational technologies. An ideal method for monitoring cellular dynamics over time should be easy to perform and should not significantly affect the properties being measured. One step towards this ideal has been the engineering of fluorescent protein variants (Giepmans *et al.* 2006). These proteins are genetically encoded and mature to functionality without requiring cofactors. Each variant fluoresces with a specific visible wavelength upon excitation, allowing multiple variants to be discerned in one cell.

Fluorescent proteins can report on the protein levels by directly creating translational fusions or indirectly creating transcriptional fusions. A translational fusion is made by inserting a fluorescent protein into the reading frame of the target protein resulting in the translation of the fluorescent protein and target protein as one molecule. That is, one can tag a target protein with a fluorescent tail. In many cases, this does not significantly affect either protein's function. A transcriptional fusion is made by co-expressing a fluorescent protein and a target protein by placing each behind the same promoter. While this strategy reports on promoter activity, a key determinant of intracellular levels, it fails to capture any post-transcriptional or post-translational regulation, such as the action of regulatory RNAs or proteases. With both transcriptional and translational fusions, fluorescence measurements are non-invasive to live cells, and the process can be automated for long-term measurements. Fluorescent proteins therefore represent an elegant solution for monitoring *in vivo* protein levels. Caution must be exercised with translational fusions, however, because even if the fluorescent tag does not alter the target protein's function *per se*, it may significantly impact its localization. Although many of such cases are unreported, the literature is spotted with examples of mis-localized or mis-transported fluorescent fusion proteins (Roucou *et al.* 2000; Hanson & Ziegler 2004). This is an important issue not only for studies that explore protein trafficking, but also for any system where altered localization will affect function.

A particularly appealing application of fluorescent proteins is to monitor single-cell dynamics in real time through optical microscopy. Single-cell measurements are critical for revealing heterogeneity in gene expression or differences in other phenotypic traits between the cells that are often masked in population-level measurements. In one of the earliest synthetic circuits published, Elowitz & Leibler built a circuit capable of producing oscillations in gene expression, but it was only through the microscopic tracking of individual lineages of bacteria that the oscillations became truly apparent (Elowitz & Leibler 2000). Similar techniques were used to characterize other oscillators implemented later (Atkinson *et al.* 2003; Fung *et al.* 2005). Recently, single-cell measurements have become the workhorse for a series of elegant

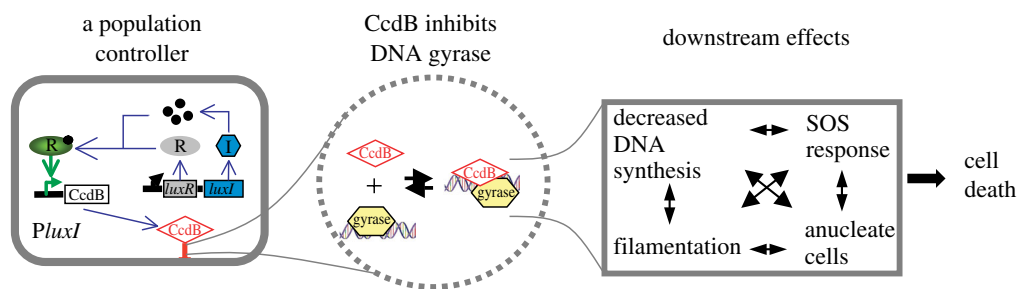


Figure 3. Complexity and uncertainty in a biological circuit design. Although we can build and model the circuit from box 1, it is remarkably difficult to capture even all the known interactions (let alone the unknown interactions). In our model, we have a single killing term that sets the rate of cell death proportional to the product of the killing rate constant, CcdB level and cell number. In reality, the situation is far more muddled. CcdB operates on DNA gyrase in a manner whose mechanistic details are still open to debate. The downstream effects of CcdB are plural and interrelated, and each of these involves many components. For example, the SOS response involves over a dozen players. Attempting to incorporate all the partially understood downstream effects would complicate the model with no guarantee of improving its accuracy. Nevertheless, by omitting them, we make the implicit assumption that they do not affect system dynamics.

experimental studies aimed at deciphering the origin and characteristics of cellular noise (Elowitz *et al.* 2002; Ozbudak *et al.* 2002; Blake *et al.* 2003; Raser & O’Shea 2004; Hooshangi *et al.* 2005; Pedraza & Van Oudenaarden 2005; Rosenfeld *et al.* 2005; Austin *et al.* 2006; Guido *et al.* 2006; Volfson *et al.* 2006).

Remarkably, measurement capabilities are continuing to improve in resolution, as tools to track single molecules *in vivo* have also been developed. Building on previous mRNA visualization techniques (Bertrand *et al.* 1998), it is now possible to track individual mRNAs *in vivo* by using multiple fluorescent mRNA-binding proteins (Fusco *et al.* 2003; Golding & Cox 2004; Shav-Tal *et al.* 2004). Yu *et al.* show that it is even possible to detect a *single* fast maturing fluorescent protein by targeting it to the membrane (Yu *et al.* 2006). These detection methods improve researchers’ abilities to quantify the abundance and localization of cellular components. Researchers can then determine when and where the experimental system deviates from their expectations, improving their ability to test and troubleshoot designs.

It is a rare and joyous occasion when a synthetic genetic circuit actually works as expected for the first time. The laborious and time-consuming process of characterizing and debugging biological programs will become more significant as the circuits increase in complexity. This process is, by and large, the rate-limiting step for engineering gene circuits that program sophisticated dynamic behaviour (box 1). An important advance in this area is the miniaturization of characterization processes through microfluidics—the science and technology of systems that manipulate small amounts of fluids (10^{-9} – 10^{-18} l), using micro-sized channels (Quake & Scherer 2000; Hong & Quake 2003). Microfluidic metering enables ultra-low consumption of biological samples and reagents, allowing high-throughput research at low cost with short analysis time. Microfluidic miniaturization also facilitates automation and integration of complex chemical or biological procedures into a single process that is faster, more precise and more reproducible than its manual counterparts. Pioneered by the Quake laboratory, the development of actuatable pneumatic valves

through multilayer soft lithography (MSL) has facilitated the design of complicated devices equipped with pumps, fluidic isolation and mixers (Unger *et al.* 2000).

As a proof of concept for synthetic biological application, Balagadde and colleagues devised and implemented a miniaturized 16 nl bioreactor, called a microchemostat, that enables automated culturing and monitoring of small populations (10^2 – 10^4) of bacteria for hundreds of hours with single-cell resolution (Balagadde *et al.* 2005). By reducing the reactor volume by a factor of 10^5 when compared with traditional chemostats, microchemostat populations undergo proportionately fewer divisions per hour, which suppress the *total* mutation rate of the population. This, in turn, effectively insulates the micro-cultures from rapid evolution, prolonging monitoring of genetically homogeneous populations. The microchemostat system is automated by the custom software that controls periodic media dilution, culture mixing, image acquisition and image analysis. Its unique design also allows multiple experiments to be run in parallel on the same chip (figure 4). In addition to the measurements of cell density and morphology, a recently improved chip design enables measurements of gene expression dynamics reported by fluorescence or luminescence (F. Balagadde, unpublished data).

In another microfluidics application, Thorsen and colleagues created a ‘comparator’ capable of screening individual cells for desired functionality in a high-throughput manner. In this device, two reagents can be separately loaded into 256 pairs of subnanolitre reaction chambers. Adjacent chambers are united allowing the reagents to mix and react. The products of each reaction can then be selectively recovered. This system was used to perform a high-throughput detection of single bacterial cells expressing recombinant cytochrome *c* peroxidase (Thorsen *et al.* 2002).

Fu and colleagues fabricated a microfluidic fluorescence-activated cell sorter (FACS) to sort live fluorescent *E. coli* cells. Compared with the conventional FACS machines, the microfluidic device allows for more sensitive optical detection of bacterial cells as well as DNA strands, and it is also capable of ‘reverse’ sorting. Reverse sorting is a procedure where cells are

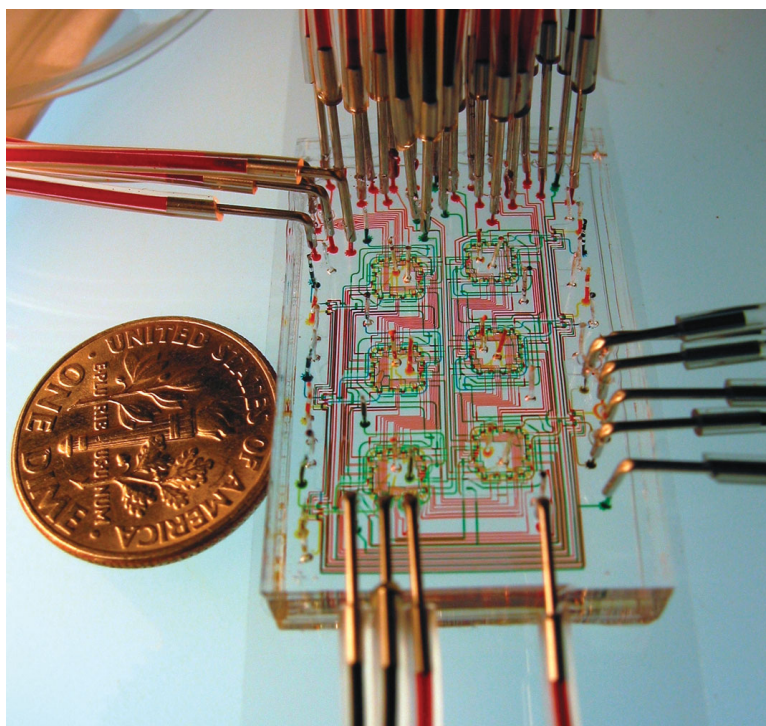


Figure 4. A microfluidic chip with six parallel microchemostat reactors, used to study the growth of microbial populations. The coin is 18 mm in diameter.

scanned at a high flow rate until a fluorescent cell is detected. Flow is then stopped and reversed, allowing the cell to be measured a second time and diverted into a collection tube. Reverse sorting is particularly useful for isolating rare cells or making multiple measurements on a single cell (Fu *et al.* 2002).

The aforementioned microfluidic devices can be used in stand-alone applications or as part of an integrated system. They are also disposable, which eliminates any cross-contamination in between the runs. These and many other microfluidic systems (Cookson *et al.* 2005; Groisman *et al.* 2005; Zhang *et al.* 2006) being actively developed will become important tools for synthetic biologists (El-Ali *et al.* 2006).

3. APPLICATIONS

3.1. Green chemistry

Natural biological systems are astonishing production factories capable of synthesizing an impressive array of chemicals with relatively high yields. For example, the plant metabolome alone is estimated to contain over 1 million unique chemical species (Schwab 2003). Furthermore, all of these diverse chemical species are synthesized under ‘gentle’ conditions in the cells (i.e. in aqueous solutions and at mild temperatures). In contrast, current methods for organic synthesis often rely upon exotic solvents, reaction conditions and catalysts. Not only are such methods expensive, but they can also produce a variety of undesirable and toxic waste products. These problems can be alleviated through development of novel biological catalysts and synthetic metabolic pathways. Such advances could usher in a new era of environmentally friendly or ‘green’

chemistry by breaking our dependence on toxic solvents and catalysts while decreasing waste product formation.

It would be naive to think that custom metabolic synthesis will replace the majority of organic chemical synthesis in the near future. However, it can have an immediate impact on several areas. One such example is the production of artemisinic acid, a precursor to the antimalarial drug artemisinin. Originally discovered as a Chinese herbal therapy, artemisinin is currently isolated from the shrub *Artemisia annua*, but it is too expensive for most populations where malaria is a problem. Total chemical synthesis is difficult and costly, but researchers have recently reported the production of up to 100 mg l⁻¹ artemisinic acid from an engineered laboratory yeast strain (Ro *et al.* 2006). To engineer the yeast strain, Keasling and colleagues first increased precursor production by manipulating the farnesyl pyrophosphate (FPP) pathway to augment FPP yield. They additionally downregulated a gene that diverts FPP to a sterol-producing pathway. They then added genes from *A. annua* to convert FPP to amorphaadiene and subsequently convert amorphaadiene to artemisinic acid. The authors report a simple purification scheme to recover the artemisinic acid, which can then be converted to artemisinin in a relatively straightforward chemical reaction. This would appear to be a vast improvement over their complementary work in *E. coli* that reported the introduction of a metabolic pathway capable of producing up to 24 mg l⁻¹ of amorphaadiene, an artemisinic acid precursor (Martin *et al.* 2003). It has recently been reported, however, that the engineered *E. coli* strain produces higher levels (500 mg ml⁻¹) than previously measured (Newman *et al.* 2006). Measurement errors were due to the high volatility of amorphaadiene in

aqueous solutions. It therefore remains to be seen which organism will ultimately be the most useful as a bioreactor for this application. That virtually the same metabolic pathway can be built in two organisms from different taxonomic domains is perhaps an indication of the potential plasticity of cellular metabolism in the hands of a skilled practitioner.

Artemisinic acid is not the first example of a therapeutic molecule produced in a cell culture. Many drugs currently on the market including insulin, erythropoietin and therapeutic antibodies are also made in cellular bioreactors. However, artemisinic acid is distinctly different from most biologically cultivated therapeutics, because it is not a protein and requires many more metabolic steps than simple transcription and translation. All these steps must be carefully regulated and balanced to control metabolic fluxes and maximize yield. This type of synthetic biology is deeply rooted in what many might call metabolic engineering (Bailey 1991; Stephanopoulos & Vallino 1991). Some may even argue that metabolic engineers have been doing synthetic biology far before the label became well established.

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3.2. Therapeutics

Drug production is only one example of how synthetic biology can contribute to medicine. In this age of shots and pills, it is easy to forget that our bodies' defence system is predominantly composed of cells. Billions of immunological cells patrol our bodies at any given time on the lookout for antigens that indicate foreign cells or abnormal function. A key feature of our immune system is that it is predominantly targeted to the particular offending pathogens or region of infection through the use of cell surface receptors and signalling molecules (Goldsby 2003). Most drugs, however, are often taken systemically and can be damaging to unintended targets. For example, many chemotherapy treatments aim to control the fast proliferating cells in the cancer, but inadvertently destroy the rapidly dividing hair follicles and cells of gastric linings, resulting in hair loss and digestion problems.

Cells can be engineered to recognize specific targets or conditions in our bodies that are not naturally recognized by the immune system. Although some drugs can also be targeted to specific locations through aptamer (Mcnamara *et al.* 2006) or antibody conjugation (Schrama *et al.* 2006), a cell has the advantage of being able to interpret and respond to complex environmental signals. Anderson *et al.* (2005) engineered bacteria to invade tumours in response to specific extracellular conditions. By directing expression of the invasion gene from *Yersinia pseudotuberculosis* through promoters responsive to hypoxia, cell density or arabinose, they restricted bacterial invasion of mammalian tumour cells to these conditions. This is significant because the tumour environment is often hypoxic and allows for high bacterial cell densities due to depressed immune function in the tumour. Therefore, this work demonstrates, as a 'proof of concept', that one can potentially use engineered bacteria to

target diseased cells without significantly impacting healthy cells.

A lot of synthetic biology research has been carried out in bacteria due to their ease of manipulation and simpler physiology when compared with mammalian cells. While engineered bacteria do have tremendous potential for therapeutic applications, as previously illustrated, the general public may feel more comfortable dealing with therapeutics derived from mammalian cells. Furthermore, mammalian cells are already closer to being optimized for functions in the human body. For these reasons, major advances in cell-based cancer therapeutics are being made through engineering of mammalian cells, including stem cells. Rosenberg and colleagues report the generation and application of tumour-specific T-cells in 15 metastatic melanoma patients (Morgan *et al.* 2006). To generate the cells, a T-cell receptor recognizing the tumour-associated antigen (TAA) MART-1 was transfected (using a retroviral vector) into the peripheral blood lymphocytes isolated from the patients. Patients received the engineered cells by adoptive cell transfer. Even though only 2 out of the 15 patients showed sustained regression, the work demonstrates the potential applicability of targeted therapy using engineered cells. The authors also indicate many possibilities for improvement in future trials, such as tighter binding TAA receptors or cytokine/tissue-homing mechanisms.

The aforementioned work used a retrovirus for integration of the transgene into patients' cells *ex vivo*. Viral vectors for gene therapy often *insert* DNA at a particular locus in the cells' chromosome. However, in many cases, it may be more desirable to actually *replace* a malfunctioning gene. Recent 'parts design' has produced a library of the so-called zinc-finger nucleases (ZFNs) that may enable *in vivo* human gene replacement. ZFNs join a type of DNA recognition element (zinc finger motifs) and a DNA-cleaving enzyme (nuclease) to target a specified sequence. Unlike many of the common bacterially isolated restriction enzymes, which recognize 4, 6 or 8 bp, ZFNs can recognize a sequence long enough for it to be unique in an organism's genome. Urnov *et al.* report the construction and application of a pair of ZFNs that recognize a 24 bp site in the human genome (Urnov *et al.* 2005). The ZFNs create a double-strand break in the chromosome. The break, in turn, induces cells' natural homologous recombination machinery to incorporate DNA from synthetic donor constructs. Twenty per cent of chromosomes successfully recombined, leading to 7% of cells homozygous for the correction in the absence of selection. ZFNs or other large and specific endonuclease design (Arnould *et al.* 2005; Ashworth *et al.* 2006) may hold the key to altering an organism's DNA post-development. Such *in vivo* genome alterations will enable therapeutic intervention ranging from simple replacement of mutant alleles with wild-type to controlled integration of novel multi-gene circuits. An intrinsic advantage of gene correction (over gene insertion) is that the replaced allele is present in its natural chromosomal locus, therefore increasing the chance it will be properly regulated.

3.3. Renewable energies

As the global supply of fossil fuels diminishes, alternative and renewable energy sources will become more critical than ever before. Production of bioethanol, ethanol derived from crops, has emerged as a potential way to convert abundant solar energy gathered by plants into easily stored fuel for combustion engines (Hahn-Hagerdal *et al.* 2006). Production of bioethanol relies upon micro-organisms, such as yeast, to ferment the plant materials. A current limitation, however, is that most naturally occurring or laboratory micro-organisms are incapable of converting all types of energy-storing compounds found in crops into ethanol. For this reason, sugarcane and corn are the major feedstocks for bioethanol conversion (both sucrose and starch can easily be converted to glucose). Consequently, bioethanol production is economically feasible only in the regions producing such crops, such as Brazil.

All human habitats have naturally thriving plants that contain other energy-storing compounds, including cellulose (40–50%), hemicellulose (25–35%) and lignin (15–20%; Grey *et al.* 2006). However, for this ‘cellulosic biomass’ to be used, improvements must be made in both the enzymatic degradation of these compounds into simpler sugars (including glucose) and the efficient conversion of non-glucose sugars to ethanol. Optimization of enzymes such as cellulose and hemicellulase can result in decreased costs and higher efficiency. Here, synthetic biology could play a role by either boosting expression through systems design or improving activity and stability through parts design.

Microbial strain engineering has already begun to tackle the issue of non-glucose sugar conversion (Jeffries 2006). The primary non-glucose sugar formed after enzymatic breakdown of cellulosic biomass is the pentose sugar xylose. As a result, early efforts have focused on the engineering of microbial strains capable of co-fermenting glucose and xylose simultaneously in order to increase yield and production rates. Ho and colleagues address the xylose utilization issue by introducing three xylose-metabolizing genes into the yeast chromosome at multiple copy numbers: xylose reductase (XR); xylitol dehydrogenase (XD) and xylulokinase (XK; Sedlak & Ho 2004). Together, these three enzymes convert xylose to xylulose-5-phosphate, a key metabolite in the yeast pentose metabolism pathway. Resulting strains produced ethanol levels in excess of 75% of the theoretical yield of sugars consumed. An alternate method, described by the Pronk laboratory, features the introduction of a fungal xylose isomerase from *Piromyces* and the over-expression of downstream pentose phosphate pathway genes: xylulokinase; ribose 5-phosphate isomerase; ribulose-5-phosphate epimerase; transketolase; and transaldolase (Kuyper *et al.* 2005). The *GRE3* gene, which produces unwanted side product xylitol, was deleted from the strain. The resulting strain was capable of fast anaerobic growth with xylose as the sole carbon source, but still showed a strong preference for glucose in mixed carbon source cultures. In subsequent work, Pronk and colleagues employed long-term nutrient-limited chemostatic cultures to

evolve strains with improved xylose uptake and usage kinetics, resulting in a strain that completely ferments both glucose and xylose in less than 25 h. While both sets of strains described here can benefit from further improvements, they demonstrate the progress being made towards expanding bioethanol production to more diverse crops. Future generations may be cultivating high yield and easy to grow species such as switchgrass or hybrid poplar trees to fuel the worlds growing energy needs.

3.4. Pattern formation

The human body is a complex system of specialized cells, tissues and organs. Remarkably, each highly specialized cell in our bodies arises from a single fertilized egg cell. This process of differentiation and morphogenesis is mediated by the delicate interplay of chemical gradients, cellular receptors, differential gene expression and cell migration (Gilbert 2000). The end result is that the 100 trillion cells of the adult human are neatly arranged and specialized in a way that allows for proper functioning of all bodily processes. Nature has produced incredibly complex systems, as well as a fantastic way of assembling them. This accomplishment is even more amazing considering that the overall robust system builds upon components that are often intrinsically ‘noisy’. With regard to this accomplishment, the synthetic biologist can ask ‘in what ways can we recreate or use the complex pattern formation systems found in nature, and to what ends?’

As a first step to address this question, Weiss and colleagues rewired cell signalling pathways to create a model system of chemical gradient-induced pattern formation in bacteria (Basu *et al.* 2005). ‘Sender cells’ produce a small membrane-diffusible chemical, acylhomoserine lactone (AHL), by expressing the *luxI* gene from *V. fischeri*. ‘Receiver cells’, in turn, respond to the signal through *luxR* activation upon AHL binding, which induces transcription from a lux promoter. By placing both a single repressor and a double repressor cascade behind lux promoters, Weiss and colleagues effectively created a band detector such that a downstream gene (*gfp*) is expressed only at intermediate concentrations of AHL. Furthermore, by creating variants of receiver plasmids through *luxR* mutagenesis and copy number reduction, receivers can be tuned to respond to different bands of AHL concentrations. Consequently, when a region of sender cells is placed within a lawn of receiver cells, fluorescence is observed only in a ring whose distance from the sender cells varies in accordance with the version of the receiver plasmid used. The visual result resembles a bullseye.

While the bullseye pattern is novel and interesting, one may be left wondering what use it can find. With a little imagination, however, one can envision using this pattern formation system to control a master regulatory gene capable of committing cells to a particular developmental fate. A higher-order function in natural biological systems is associated with multi-cellularity and cellular specialization. To produce similarly complex functions, synthetic biologists will require mechanisms that produce and maintain differentiation

1387 patterns. These mechanisms may lead to highly
 1388 sophisticated cellular system for fabricating biomater-
 1389 ials with well-defined dimensions. This line of research
 1390 may also synergize with research efforts focusing on
 1391 regenerative medicine (Lagasse *et al.* 2001) and tissue
 1392 engineering (Griffith & Naughton 2002), both of which
 1393 hinge upon controlling differentiation and pattern
 1394 formation. Biologists' continued efforts to implement
 1395 synthetic multi-cellular systems will drive the pro-
 1396 duction of new and better approaches to artificial
 1397 cellular communication. Most communication systems
 1398 employed by synthetic biologists thus far have made
 1399 use of the small diffusible molecules from bacterial
 1400 quorum sensing. Further developments may feature
 1401 active and regulated transport of signalling molecules
 1402 across the cell membrane and the use of cell surface
 1403 receptors to recognize and send signals to adjacent cells.
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1406 4. OUTLOOK

1407 4.1. Standardization: promises and limitations

1408 It has been suggested that many of the difficulties in the
 1409 production and optimization of biological circuits are
 1410 due to improper and incomplete description of parts
 1411 (Endy 2005). These limitations are twofold: first,
 1412 functional characteristics are often unknown for many
 1413 parts; second, even if they are known, they are rarely
 1414 described using standardized measures and are often
 1415 buried in the literature. Towards addressing these
 1416 limitations, the BioBricks Foundation has established a
 1417 'registry of standard biological parts' (The_BioBricks_
 1418 Foundation). The registry categorizes parts, devices
 1419 and systems. Ultimately, the registry strives to provide
 1420 information on not only sequence but also functional
 1421 characteristics, and make information available
 1422 through a central portal. Many of these parts have
 1423 been cloned into plasmids that enable easy assembly.
 1424 The plasmids are made available to students partici-
 1425 pating in the international Genetically Engineered
 1426 Machine competition (iGEM). Members of the Bio-
 1427 Bricks Foundation hope that the registry will decrease
 1428 the time and research costs needed to design and
 1429 implement gene circuits. Such efforts are analogous in
 1430 spirit to ongoing attempts to standardize mathematical
 1431 models (Hucka *et al.* 2003) and formats for microarray
 1432 data (Brazma *et al.* 2001). The limits in achieving parts
 1433 standardization for *E. coli* and other organisms remain
 1434 to be seen.
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1436 Even with a repository of information about stan-
 1437 dardized parts, a major challenge to applying this
 1438 information will be developing strategies to deal with
 1439 context dependence (Andrianantoandro *et al.* 2006;
 1440 Arkin & Fletcher 2006). For example, synthetic gene
 1441 circuits often exhibit varying behaviour in different cell
 1442 strains. In some cases, this can be easy to rationalize by
 1443 the presence or absence of a particular gene, or a
 1444 documented difference in the growth rate. In other cases,
 1445 causes of variability are much more difficult to ascribe
 1446 due to many hidden interactions between the designed
 1447 circuit and a far-from-elucidated host circuitry.
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1449 To address this issue, one may imagine selecting a
 standard cell strain, in which standard parts under

standard conditions are to be quantified. A starting
 point for such a standard strain may be on its way. The
 Blattner group has recently engineered a series of
 multiple deletion strains (MDS) that have up to 15% of
 their parental MG1655 genome removed but maintain
 similar growth rates on minimal media (Posfai *et al.*
 2006). Deletions were guided by comparative genomics
 with related strains. Removing 'unnecessary' portions
 of the genome can presumably reduce the number of
 hidden interactions. Notably, the deletions cleaned the
 cells of mobile DNA elements called insertion sequences
 (IS) that might reduce the genetic stability of a circuit
 by inserting themselves into and disrupting a DNA
 sequence unpredictably. Interestingly, the MDS strains
 produced some unanticipated benefits, including higher
 electroporation efficiencies than their parent strain and
 the ability to propagate some plasmids that the parent
 strain could not.
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1468 In an alternate approach, researchers at the Venter
 Institute have used *Mycoplasma genitalium* as a
 starting point in their attempts to determine a minimal
 gene set by systematically mutating every gene (Glass
et al. 2006). *Mycoplasma genitalium* has the smallest
 known genome that is capable of growth in the absence
 of other species. They conclude that in a laboratory
 setting, only 382 of the strain's 482 genes are essential.
 Although a strain containing only this set of minimal
 genes has not yet been constructed, it could eventually
 serve as a bare bones platform upon which desired
 functionality can be added. Such a small number of
 genes might allow a greater percentage of the cell's
 molecular interactions and metabolic processes to be
 understood, making the strain more predictable and
 desirable as a starting point. However, of the 382
 essential genes determined, 110 are annotated as
 hypothetical proteins or as proteins of unknown
 function, indicating that a truly complete cellular
 model, even for this simplest of cells, cannot yet be
 produced.
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1489 Despite characterizing parts in a standard strain
 under defined conditions, individual parts may impact
 the physiology of the host strain differently, for
 instance, by placing varying burdens on the host
 translation machinery. For this reason, one may wish
 to minimize such interactions by creating privileged
 sets of machinery. For instance, Rackham & Chin
 (2005b) describe the formation of orthogonal ribo-
 some—mRNA pairs that could be used to keep a
 synthetic system and host more isolated. Using a dual
 positive–negative selection scheme, they isolated
 mRNAs with modified Shine–Dalgarno regions not
 recognized by endogenous ribosomes, but instead
 recognized by alternative ribosomes. Translation by
 orthogonal pairs should be unaffected by endogenous
 ribosomes and there should be no competition for
 ribosomes between orthogonal mRNAs and traditional
 mRNAs. In principle, multiple ribosome types can be
 implemented for a specified function, just as cells
 already possess multiple DNA or RNA polymerase
 types, which play specialized roles.
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1510 Previous and current progress promises an ever
 growing infrastructure that will no doubt tremendously
 benefit future synthetic biology research, fundamental
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1513 and applied alike. Concerning standardization,
1514 however, two critical questions remain to be addressed
1515 by the community. First, given the amount of cell
1516 physiology (even for highly characterized organisms
1517 such as *E. coli*) that is still poorly understood, to what
1518 extent can we standardize parts or systems with
1519 confidence? Second, how much standardization can
1520 we afford and still hope to create useful systems that
1521 can work in complex environments such as in a cancer
1522 or a polluted environment?

1523 There is little difficulty in unambiguously defining the
1524 DNA sequences that code for parts, be they proteins or
1525 RNAs. The true challenge lies at the functional levels.
1526 Parts will impact and be impacted by cell physiology,
1527 which also changes in response to the environmental
1528 conditions. In addition, parts tested in isolation may
1529 unpredictably impact each other's functions when
1530 combined. For example, connecting one part's DNA
1531 with another part's may introduce unintended
1532 regulation by introducing enough flexibility in DNA to
1533 allow DNA looping. For these reasons, one can rarely
1534 have complete confidence in the part's function even if
1535 he/she uses it in a standard strain characterized under a
1536 standard condition. Many such interactions are still
1537 poorly understood, complicating the use of standard
1538 parts. Yet, it is precisely this complexity that makes
1539 engineering biology challenging and interesting. Decod-
1540 ing this complexity is at least one important application
1541 of synthetic gene circuits. Without a much deeper
1542 understanding of cellular functions at all levels, it is
1543 difficult to even define standards meaningfully.

1544 From a practical standpoint, too much standard-
1545 ization may remove flexibility in engineering useful
1546 systems. It would be illogical to rely only on standard
1547 strains that lack desirable properties for a particular
1548 application. Consider thermophilic bacteria, capable of
1549 life at temperatures as high as 113°C (Stetter 1999).
1550 The ability to thrive at elevated temperatures may be a
1551 useful property for synthetic organisms involved in
1552 chemical processing, because higher temperatures
1553 speed kinetic rates. Given the difficulty in thermo-
1554 stabilizing even a single protein, however, it is unlikely
1555 this quality can be engineered into a standard strain.
1556 For many applications, the researcher is left with no
1557 appealing options except to use non-standard strains.
1558 No single strain or growth condition can ever cover all
1559 potential synthetic biology applications.

1560 If we remain dedicated to standardization, gathering
1561 standardized information for a set of potentially useful
1562 parts, in a set of useful strains, under a set of relevant
1563 conditions becomes a combinatorial nightmare. The
1564 inevitable result is that standards will only be available
1565 for a limited number of strains and conditions.
1566 Although some information is preferable to none, a
1567 rising danger is to place undue weight on the limited
1568 information available and assume that a part's
1569 behaviour will not vary significantly from the context
1570 in which it was described. In this situation, 'significant'
1571 is considered to be variation that exceeds the accep-
1572 table tolerance limits of a part in its new device.
1573 Accepting standardized information at face value,
1574 without acknowledging its limitations, will lead one to
1575 design many systems doomed to fail. However, being

1576 aware of the limitations allows one to use standard
1577 information without depending on it, to be guided by
1578 the information while simultaneously embracing
1579 strategies like combinatorial design (Guet *et al.* 2002)
1580 and directed evolution (Yokobayashi *et al.* 2002) of
1581 circuits—strategies that would be unnecessary in a fully
1582 standardized and predictable world.

1583 4.2. *De novo cells*

1584 Finally, synthetic biology may, in addition to redesign-
1585 ing cellular processes, contribute to producing artificial
1586 cells exhibiting all the qualities that we associate with
1587 life. For a good review of what characteristics such a cell
1588 would need and what progress has already been made,
1589 see Deamer (2005). It is probable that no matter what
1590 system is devised for artificial encapsulation of
1591 materials in membranes capable of self-reproduction,
1592 there will be argument as to whether life has truly been
1593 created. In fact, somatic cell nuclear transfer, best
1594 known for cloning Dolly, the sheep, has already
1595 accomplished a cellular 'cold boot'. At the moment
1596 that the nucleus containing the DNA (software) is
1597 removed from the somatic cell, it is no longer living by
1598 standard definitions and could be considered a collec-
1599 tion of nucleic acid and protein molecules. Similarly, an
1600 enucleated egg cell and cytoplasm (hardware) is not
1601 alive by consensus definitions. However, when the two
1602 are combined, life arises anew. The immediate retort
1603 might be that the system relies too heavily on cellular-
1604 derived components. Where does one draw the line
1605 however? Will it only officially be the 'creation of life' if
1606 each protein, nucleic acid or lipid in the new pseudo-
1607 cells is chemically synthesized from precursors? Will
1608 precursors themselves need to be produced from pure
1609 elements? In any case, de novo cell design can shed light
1610 on both the properties needed to produce life and how
1611 terrestrial life could have arisen initially.

1612 4.3. *Social impact*

1613 Synthetic biology will undoubtedly head in many
1614 unforeseen directions in the coming years and decades,
1615 but along the way, researchers in the field are paying
1616 particular attention to legal, ethical and political issues
1617 dealing with the redesign of life. At the second annual
1618 Synthetic Biology conference (SB2.0) in 2006, a full
1619 third of the time was devoted to issues of bio-safety,
1620 public perception, ownership and community organiz-
1621 ation. Even in the early stages of the field, the need for
1622 this discussion was apparent to many. Although
1623 screening and controls have been put in place by
1624 many DNA synthesis companies, these technologies can
1625 allow for the purchase of potentially dangerous genetic
1626 material. Nothing has illustrated this point more
1627 clearly than the 2002 production of poliovirus from
1628 synthetic DNA by the Wimmer laboratory (Cello *et al.*
1629 2002). Using only synthetic oligos, cells expressing T7
1630 polymerase and cell-free extracts, an 'eradicated' virus
1631 with the same pathogenic properties as the original was
1632 reproduced. Public perception issues were highlighted
1633 by the publication of an open letter from a group of
1634 NGOs, including Greenpeace and ETC that called for
1635

1639 synthetic biologists to drop plans for self-governance
1640 and instead demand governmental supervision due to
1641 the ‘potential power and scope of this field’.

1642 A consideration of ownership issues arises from the fact
1643 that if biological parts are owned and protected by
1644 various entities, it may be legally difficult to produce a
1645 complex system incorporating many of those parts—
1646 hindering innovation and potential societal good. Active
1647 analysis by legal scholars is needed to develop systems
1648 that ensure freedom to operate, but maintain incentives
1649 for invention and development. Some of this analysis is
1650 already underway (Rai & Boyle *forthcoming*). Finally,
1651 the need for community organization is evident in order to
1652 not only manage the issues of public perception, bio-
1653 safety and ownership, but also to guide the field in a way
1654 that reduces growing pains. Particularly important is the
1655 prevention of unrealistic expectations on the part of
1656 granting agencies and public. Synthetic biology holds a
1657 lot of promise, but none of the fields can address all
1658 problems and none have produced any answers over-
1659 night—despite popular hype. To achieve its vast
1660 potential, synthetic biology will need sustained support
1661 from governments and a public that understands progress
1662 is made in incidental steps.

1665 7. UNCITED REFERENCES

1666 Rosenfeld *et al.* (2002) and Kobayashi *et al.* (2004).

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2143	Author Queries	2206
2144	<i>JOB NUMBER:</i> 20060206	2207
2145		2208
2146	<i>JOURNAL:</i> RSIF	2209
2147		2210
2148	Q1 Please check the sense of the sentence ‘For an	2211
2149	overview of where these technologies interact with	2212
2150	synthetic biology (figure 1)’. Would this be better to	2213
2151	read as ‘figure 1 illustrates an overview of where these	2214
2152	technologies interact with synthetic biology’?	2215
2153	Q2 Please provide the year for the reference ‘J. Tian	2216
2154	(personal communication)’.	2217
2155	Q3 Please approve the edit of the sentence ‘Up a level of	2218
2156	abstraction from the DNA synthesis...’ to ‘One level of	2219
2157	abstraction from the DNA synthesis...’	2220
2158	Q4 Please note the edit of the term ‘effector’ to ‘effector’ in	2221
2159	the sentence ‘RNA switches are advantageous in their	2222
2160	fast response...’	2223
2161	Q5 Please clarify ‘custom make parts’ in the sentence	2224
2162	‘These examples demonstrate that the cellular engin-	2225
2163	eer...’	2226
2164	Q6 Please check whether ‘behaviours’ should be	2227
2165	‘behaviour’ in the line ‘Such analyses are useful to	2228
2166	determine how quantitative or qualitative circuit	2229
2167	behaviours...’	2230
2168	Q7 In the sentence ‘In many cases, this does not	2231
2169	significantly affect either protein’s function’, please	2232
2170	confirm whether this would be better to read as ‘In	2233
2171	many cases, this does not significantly affect the	2234
2172	function of either a target or a fluorescent protein’?	2235
2173	Q8 Please provide the year for the reference ‘F. Bala-	2236
2174	gadde (unpublished data)’.	2237
2175	Q9 Please approve the edit of the sentence ‘...one may	2238
2176	imagine selecting a standard cell strain in which to	2239
2177	quantify standard parts under standard conditions’ to	2240
2178	‘...one may imagine selecting a standard cell strain, in	2241
2179	which standard parts under standard conditions are to	2242
2180	be quantified’.	2243
2181	Q10 References Rosenfeld <i>et al.</i> (2002) and Kobayashi <i>et</i>	2244
2182	<i>al.</i> (2004) are provided in the list but not cited in the	2245
2183	text. Please supply citation details or delete the	2246
2184	references from the reference list.	2247
2185	Q11 Please check the year, volume number and page	2248
2186	range for the reference ‘Arnould <i>et al.</i> (2005)’.	2249
2187	Q12 Please update the year for the reference ‘Rai & Boyle	2250
2188	(forthcoming)’.	2251
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