

Non-model bacteria as platforms for endogenous gene expression in synthetic biology

Jorien Poppeliers^{1,2}, Maarten Boon¹, Marjan De Mey³, Joleen Masschelein^{1,2,4} & Rob Lavigne¹✉

Abstract

Microbial synthetic biology seeks to engineer bacterial genomes for industrial and biomedical applications, typically by applying heterologous gene expression in well-characterized model organisms, such as *Escherichia coli*. However, heterologous gene expression might cause metabolic disruptions, thereby impacting production efficiency and yield. In this Review, we highlight non-model organisms, such as *Lactocaseibacillus* and pseudomonads, for endogenous compound production, taking advantage of their evolutionary optimization for the production of certain metabolites and proteins. We first outline key limitations of heterologous production and then examine endogenous production pathways in non-model organisms for biotechnological and therapeutic applications. In particular, multi-omics approaches enable the discovery and characterization of these organisms, and phage-based genome refactoring enhances genome engineering capabilities. Finally, we outline key bottlenecks in the application of non-model organisms in biotechnology, including scale-up, costs and safety.

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¹Laboratory of Gene Technology, Department of Biosystems, KU Leuven, Leuven, Belgium. ²Laboratory for Biomolecular Discovery and Engineering, Department of Biology, KU Leuven, Leuven, Belgium. ³Metabolic Engineering of Microorganisms group, Department of Biotechnology, UGent, Ghent, Belgium. ⁴Center for Microbiology, VIB-KU Leuven, Leuven, Belgium. ✉e-mail: rob.lavigne@kuleuven.be

Key points

- Chassis diversification is essential to unlock the full potential of microbial synthetic biology.
- Native producers have an evolutionary advantage over heterologous expression hosts, which can be leveraged for biotechnological applications.
- Multi-omics techniques enable the characterization of non-model organisms to become the next generation of synthetic biology chassis.
- Bacteriophage genomes are valuable sources for the discovery of synthetic biology tools to modify gene expression in non-model organisms.
- The switch towards endogenous gene expression can improve market adoption for the second generation of microbial cell factories.

Introduction

In microbial synthetic biology, genomes of microorganisms (that is, bacteria or fungi) are typically engineered to customize or create new biological processes¹. In particular, applications in industry and biomedicine have emerged by recreating, mimicking or modifying microbial metabolic processes and exploiting molecular cloning and omics technologies². Basic biological circuits can be designed to mimic logic behaviour, and entire biosynthetic pathways can be constructed to engineer synthetic cells^{3–5}. Such engineered microorganisms can be harnessed as sustainable factories for the production of ‘green’ chemicals (for example, biofuels, surfactants and herbicides) or they can be transformed into fuel cells to generate renewable energy^{6–10}. In addition, microorganisms can be designed to function as live-attenuated vaccines or probiotics^{11,12}.

In microbial synthetic biology, the term chassis is typically used as a metaphor for the framework or foundation upon which genetic engineering is built¹³. In particular, model organisms, such as *Escherichia coli*, have served as chassis in microbial synthetic biology^{1,14}. Owing to their well-characterized metabolism and regulatory networks, the modification of their genome and the introduction of new genetic material is facilitated. Therefore, outcomes of genetic modifications can be anticipated, reducing the iterative process required to obtain an optimal host for the target application. Moreover, engineering protocols and DNA delivery tools are well established for model organisms^{15,16} to serve a variety of applications^{17–21}. Accordingly, many biotechnological applications are biased towards these model organisms, even if these demands often push them beyond their natural capacity.

An alternative approach is to select the host organism best suited for a given condition and application²². Application-oriented synthetic biology requires engineered chassis to function in industrially relevant environments¹⁴, which are often characterized by alternative feedstock, such as waste streams, limited aeration and fluctuations in temperature or pH^{23,24}. However, the performance of model organisms can be affected by environmental stresses, and therefore, their potential as fermentation hosts remains limited. Alternatively, bacteria, including *Lactocaseibacillus*, *Burkholderia*, *Streptomyces*, *Bacillus subtilis* and *Pseudomonas putida*, have been explored as chassis, as they are

inherently better adapted to industrial conditions owing to the harsh environments in which they naturally occur^{13,14}.

Through evolution, many bacteria have naturally developed specialized genes or biosynthetic gene clusters that enable them to thrive in distinct environments. Their metabolic pathways are thus inherently optimized to produce certain compounds, as they already possess all the necessary resources, such as precursor molecules, antitoxins and protein-folding chaperones. Using these endogenous systems avoids the need to introduce large fragments of foreign DNA, thereby minimizing the metabolic burden associated with resource reallocation²⁵. This makes endogenous gene expression, especially when combined with genome engineering, a powerful approach for leveraging non-model bacteria in biotechnological applications^{22,26}.

In this Review, we discuss non-model bacteria as native production hosts for applications in microbial synthetic biology. We first outline shortcomings of heterologous expression in model organisms and the evolutionary advantages that favour native producers for specific applications. We then provide a framework for the implementation of non-model bacteria in biotechnology. In particular, we highlight a multi-omics approach to guide engineering as well as to identify and address potential bottlenecks, and examine phage genomes as resources to extend the synthetic biology toolbox beyond model organisms. Finally, we give an overview of current and potential future industrial applications that use non-model organisms.

Limitations of model organisms for heterologous production

Heterologous gene expression requires the introduction, transcription and translation of foreign DNA to produce a functional output (for example, protein or metabolite). This approach is typically used if genetic manipulation or cultivation of the native host is challenging^{19,27}. It has been applied for the production of human insulin, interferon- α 2b, flavour compounds and biofuels^{28–31} (Fig. 1). However, expressing genes outside their native biological context can result in reduced production yields, compared with the original producer³².

Codon usage bias

A key challenge for heterologous gene expression is codon usage bias; here, codons of foreign genes may not match the preference of the new host. This can be mitigated by silent mutations, introduced through gene synthesis³³ or codon-randomized refactoring of biosynthetic gene clusters³⁴. However, specific codon usage is considered a fine-tuned regulatory mechanism affecting both transcription and translation. Codon choice impacts posttranscriptional and posttranslational modifications, influencing mRNA levels as well as the structure and function of proteins³⁵. In addition to reliance on gene synthesis, costly and labour-intensive screening is often needed to identify improved pathways. Alternatively, rare transfer RNA molecules can be coexpressed to reduce the demand of native transfer RNAs³⁶. However, this approach can disrupt cellular resource allocation and impose fitness costs on engineered cells³⁶.

Integration of DNA

Synthesized DNA fragments must be cloned into a new host. In vitro assembly methods, including Gibson assembly and Golden Gate cloning, enable the integration of multiple DNA fragments into a plasmid backbone^{37,38}. However, biosynthetic gene cluster fragmentation can disrupt the native transcriptional architecture and, consequently, decrease product yields. Direct pathway cloning, combining long

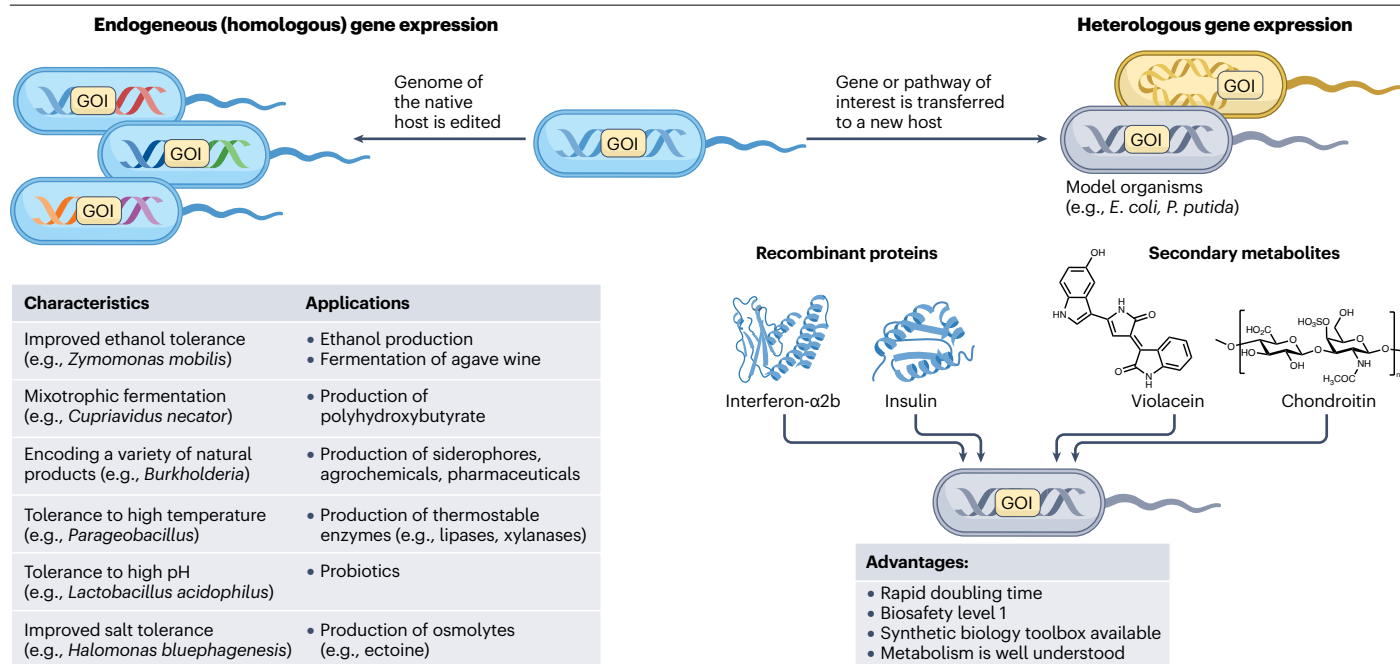


Fig. 1 | Heterologous and endogenous gene expression in biotechnology. In endogenous, or homologous, gene expression, the genome of the native host is edited. Evolution has selected native producers to thrive under specific environmental conditions, making them valuable for diverse biotechnological applications. However, a synthetic toolbox to engineer their genomes is lacking.

Consequently, heterologous expression in model organisms is frequently chosen as an alternative strategy. Here, the gene or biosynthetic gene cluster of interest (GOI) is transferred to a model organism to produce the desired protein or metabolite^{218–222}. *E. coli*, *Escherichia coli*; *P. putida*, *Pseudomonas putida*.

amplification polymerase chain reaction with in vitro recombination, addresses this limitation by enabling the cloning of entire pathways into a plasmid backbone³⁹. However, amplifying and cloning biosynthetic gene clusters of large size remains complex. Alternatively, artificial chromosomes can accommodate DNA fragments of up to 300 kb and can thus maintain and express large biosynthetic gene clusters^{40,41}. Transformation-associated recombination, which is an in vivo DNA assembly method that relies on artificial chromosomes to express entire biosynthetic gene clusters^{42,43}, has been applied in *Streptomyces* to express the tautomycin cluster (80 kb), which produces a potential anticancer molecule⁴⁴, and several aromatic polyketides (biosynthetic gene clusters of up to 213 kb) with antibacterial properties⁴⁵. Although predominantly used in *Streptomyces*, artificial chromosomes have also been applied in other hosts, including *Pseudomonas*, *Salmonella* and *Agrobacterium*^{46,47}. Nevertheless, constructing artificial chromosomes remains time-consuming and technically challenging, especially for GC-rich sequences, and does not guarantee heterologous expression⁴³. In addition, these vectors typically occur in low copy numbers, which limits expression levels. Once assembled, the DNA needs to be transferred to the host of interest, typically by chemical transformation, electroporation or conjugation. Although well studied for model organisms, the efficiency of these methods is strain dependent, and implementation in newly discovered species can be difficult⁴⁸.

Disruption of metabolic flux

The introduction of new DNA fragments can also disrupt the metabolic flux of the host organism. Therefore, reallocation of cellular resources is required to express additional genes or biosynthetic gene clusters¹⁸.

For example, the metabolic flux of *E. coli* can be reprogrammed to optimize production of chondroitin, a glycosaminoglycan that is used as a supplement to treat osteoarthritis⁴⁹, by introducing three heterologous enzymes; however, this requires changes in medium composition (that is, supplementation of the precursor) and in several metabolic pathways to attain relevant production levels⁵⁰. Therefore, additional engineering is required to alter the metabolic flux in favour of the heterologous compound. Furthermore, the absence of certain resources, such as RNA-binding proteins to stabilize mRNA, chaperones to guide protein folding or export signals to avoid accumulation in the cell and facilitate downstream processing, can impede recombinant protein production, increasing both labour and cost^{19,21,51}. These challenges in heterologous gene expression remain to be addressed.

Endogenous gene expression in non-model organisms

A second approach to alter gene expression in bacteria is endogenous (or homologous) expression (Fig. 1). Endogenous expression can be applied only if the coding sequence or biosynthetic gene cluster of interest is natively present in the chassis⁵². With over 10⁵ bacterial species described so far, there is potential to diversify the range of chassis strains available for synthetic biology⁵³.

Evolutionary adaptation and fitness landscapes

Through evolution, bacterial populations have adapted to survive and function in specific ecological niches. The concept of fitness landscapes can be applied to describe how evolutionary pressure has shaped such adaptations⁵⁴ by mapping the genotype–phenotype

relationship as a 3D landscape with bacterial populations navigating peaks and valleys in response to environmental changes. Fitness landscapes change according to the dynamics of the environment, causing new peaks to emerge that drive the evolutionary trajectory of bacterial populations. This fine-tuned, dynamic interplay between genetic traits and environmental factors determines both bacterial fitness and phenotypic traits endogenously expressed by bacteria^{54,55}. For example, *Pseudomonas* species have evolved to colonize different niches: *Pseudomonas aeruginosa* colonizes the human lungs, whereas *Pseudomonas syringae* and *Pseudomonas fluorescens* inhabit plants and soil, respectively. Consequently, distinct characteristics have emerged among these species, ranging from human and plant pathogenicity (for example, *P. aeruginosa* and *P. syringae*) to plant growth promotion (for example, *P. fluorescens*)^{56–58}.

Rapidly altering genetic (for example, through introduction of a new gene or pathway) or environmental (for example, laboratory-scale to industrial-scale) conditions can result in the maladaptation of engineered populations to the applied changes^{59,60}. For example, the expression of green fluorescent protein (GFP) can serve as an indication for the stability of synthetic circuit expression in *E. coli* over multiple generations. Upon GFP expression from two circuits, induced by either *N*-acetylhomoserine lacton or isopropyl β -D-1-thiogalactopyranoside), *E. coli* loses its ability to produce GFP after 20–50 generations, probably owing to the lack of selective pressure⁶¹. Therefore, gene–environment interactions should be considered as a coupled system, rather than as isolated variables. Moreover, it remains unclear whether new gene–environment combinations fail because they have not been favoured by evolutionary pathways shaped by past genetic changes, or because such combinations are inherently disadvantageous and, thus, selected against⁶².

Applications of extremophiles

Instead of altering multiple variables in model bacteria, chassis can be selected that require limited genetic alterations (for example, promoter exchange) or that can thrive in environmental conditions that resemble the original habitat. For example, extremophiles can be applied in processes that require operation under harsh environmental conditions, such as high temperature, high pH, high salinity or the presence of toxic compounds (Fig. 1). Extremophiles encode specialized pathways or enzymes that function under these conditions⁶³. For example, lipases and xylanases derived from thermophiles are commonly used in laundry detergents or for the degradation of paper waste, respectively^{64,65}. Extremophiles also produce a range of metabolites that can be used for biomanufacturing. For example, halophiles produce the natural product ectoin, which is used in skin care products⁶⁶. Furthermore, they are capable of producing high amounts of polyhydroxyalkanoates, which are biodegradable plastics with numerous applications in medicine, agriculture and packaging⁶⁷. Beyond biomanufacturing, acidophiles, such as *Lactobacillus acidophilus*, are used as probiotics. Their tolerance for acidic environments allows them to produce metabolites that lower the pH, preventing gastrointestinal infections and reducing serum cholesterol levels⁶⁸. In addition, the extremophilic bacterium *Bhargavaea beijingensis* has been applied for bioremediation and the production of bio-cement⁶⁹.

Biotechnological and therapeutic uses of mesophilic bacteria

Native bacteria can also be applied in vaccines to protect against bacterial pathogens⁷⁰. For example, the EvoVax vaccine, an oral vaccine consisting of four peracetic acid-inactivated *Salmonella* strains, is being

tested in piglets to treat salmonellosis, and the *Bacillus Calmette–Guérin* vaccine, a live-attenuated strain of *Mycobacterium bovis*, is approved to prevent tuberculosis^{70,71}. In addition, mesophilic bacteria can be used for biomanufacturing. For example, *Cupravidus necator* is well suited for the production of biopolymers, including poly-3-hydroxybutyrate, owing to its ability to switch between a heterotrophic and autotrophic lifestyle. *Burkholderia* spp. have large genomes, in which they encode a high number of natural products, such as pharmaceuticals, lipopeptides and siderophores. Moreover, *Zymomonas mobilis* can produce and tolerate high levels of ethanol^{72–74} (Fig. 1).

Screening by multi-omics approaches

Microbial chassis strains can be selected on the basis of their natural specializations to fully leverage endogenous gene expression in synthetic biology^{26,75}. However, although non-model bacteria have an evolutionary advantage, the mechanisms underlying their adaptations need to be understood to select the most suitable chassis for a given biotechnological application. Many bacteria have been isolated for specific purposes, resulting in limited characterization beyond their initial application⁷⁶. However, these underexplored strains may hold untapped potential for biotechnological applications.

Multi-omics approaches provide powerful tools for the identification, screening and selection of suitable microbial hosts for endogenous gene expression by offering insights into the working mechanisms of microorganisms (Fig. 2). Following isolation of a strain, genomics enables the identification of the bacterial strain and the compounds it encodes. However, genome annotations can be incomplete, necessitating additional experimental procedures, such as transcriptomics, proteomics and metabolomics, to refine annotations of regulatory elements, biosynthetic gene clusters and hypothetical proteins. Regulatory and metabolic bottlenecks can then be identified through a combination of transcriptomics, ribosomal profiling and fluxomics. Such multi-omics information can be integrated in a genome-scale metabolic model to simulate the genotype-to-phenotype relationship. This system-level understanding supports rational engineering of non-model bacteria into optimized microbial cell factories.

Genomics

DNA sequencing and genomics are key pillars in the identification and characterization of microorganisms. Currently, 715,230 bacterial genomes are publicly available in databases⁵³, and many more remain to be discovered. Therefore, identifying suitable chassis can be challenging. To capture the potential of bacteria to perform a certain function, accurate genome annotations are crucial, matching DNA sequences to their potential function. For example, bacterial genome annotation tools, such as prokka and bakta^{77,78}, perform annotation based on sequence similarity to proteins uploaded in databases, such as RefSeq and Uniprot^{79,80}. Therefore, genome annotation is limited to proteins that have been (experimentally) characterized⁸¹. As the 3D structure of a protein is intrinsically related to its function, the development of open-source software to computationally predict protein structures (for example, ColabFold and RoseTTAFold) has greatly increased the number of available protein structures and their functions in databases^{82,83}. To further optimize annotations, a multitude of tools can be applied for specific queries, including for biosynthetic gene clusters (for example, antiSMASH and MiBIG), genes related to the bacterial immune system (for example, DeepDefense) and mobile genetic elements (for example, geNomad and PHASTEST)^{84–88}. Nevertheless, the accurate prediction and annotation of small open reading

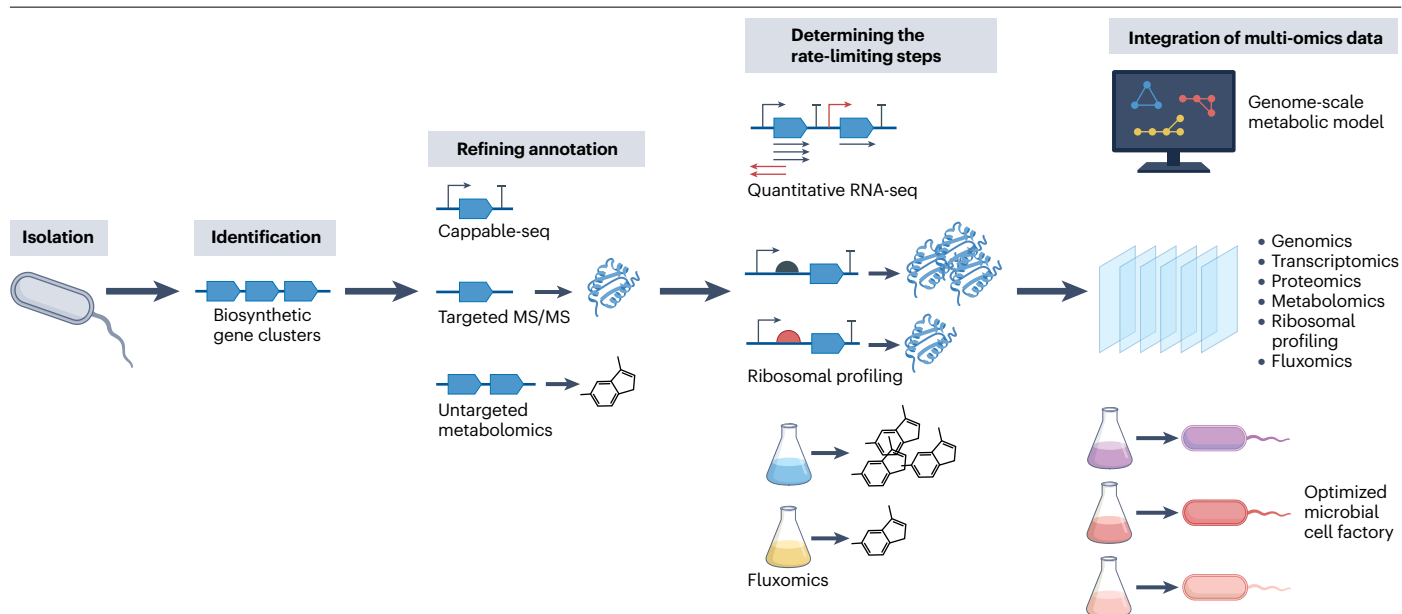


Fig. 2 | Omics technologies guiding the transition from isolation to microbial cell factories. Non-model bacteria can be developed into microbial cell factories using a multi-omics approach. Following isolation, genomics can aid in the identification of bacteria and biosynthetic gene clusters. However, genome annotations are not complete, and thus, additional experimental procedures (that is, transcriptomics, proteomics and metabolomics) can be applied to refine annotations with regards to regulatory elements, biosynthetic gene clusters

and confirmation of hypothetical proteins. Once the compound of interest is identified, regulatory and metabolic bottlenecks can be identified through transcriptomics, ribosomal profiling and fluxomics. Finally, omics information can be integrated in a genome-scale metabolic model to simulate the genotype-to-phenotype relationship. MS/MS, tandem mass spectrometry; RNA-seq, RNA sequencing.

frames (for example, peptides), non-canonical proteins and regulatory sequences (for example, promoters, terminators, ribosomal binding sites and small RNAs) remains difficult, owing to the lack of experimental validation, limiting the use of genomics alone to select a suitable production host.

Transcriptomics

Understanding how production is regulated is key to overcome bottlenecks in production and identify strategies that target such bottlenecks. For example, transcriptomics and fluxomics data revealed that heterologous expression of the cellulase CelA in *Streptomyces lividans* resulted in upregulation of genes involved in secretion, making the secretion system a potential bottleneck that can be targeted for strain improvement⁸⁹. Computational tools that can help in the identification of regulatory elements are typically based on experimental data from a limited number of model organisms. Therefore, extrapolating these results to non-model organisms might result in false positives⁹⁰. Consequently, experimental approaches are required to elucidate the regulatory mechanisms of non-model bacteria.

At the transcriptome level, RNA sequencing (RNA-seq) provides information about gene expression under a set of prespecified conditions⁹¹. For example, RNA-seq was used to determine the genes involved in biofilm formation of *Streptococcus suis* to reveal its pathogenesis and identify potential drug targets⁹². Furthermore, the transcriptome can provide information on the location of regulatory elements (that is, promoters and terminators). However, RNA-seq can typically not differentiate between processing sites and transcriptional boundaries (that is, transcriptional start and termination sites).

Therefore, experimental approaches, such as cappable-seq and terminator 5'-phosphate-dependent exonuclease treatment, are used^{93,94} to target the characteristic 5' triphosphate group at the transcriptional start site and accurately identify its position. Nevertheless, the identification of transcriptional termination sites and transcriptional units remains difficult owing to the short reads, which cannot capture full-length transcripts. To overcome this limitation, cappable-sequencing methodology can be combined with long-read sequencing technologies (that is, single molecule, real-time (SMRT)-cappable-seq and Oxford Nanopore Technologies (ONT)-cappable-seq)^{95,96}. However, although long reads might enable the sequencing of full-length transcripts, degradation at the 3' end limits the accurate identification of transcriptional termination sites⁹⁷.

RNA-seq can also be applied for identifying non-coding RNAs, which can occur as free-floating species that bind to genes or regulatory elements to fine-tune expression. Furthermore, they can act as riboswitches in the untranslated region of transcripts, where they form secondary structures that can interfere with transcription or translation⁹⁸.

Proteomics and metabolomics

Proteomics and metabolomics rely on mass spectrometry to identify compounds and biosynthesis pathways. In addition, reverse genetics approaches, such as gene knockout screens, can help link the genome to the proteome and metabolome⁹⁹. In contrast to the DNA sequence, which typically remains stable over time, the protein and metabolite content of a cell greatly depend on the conditions under which the bacteria are analysed. For example, studying the proteome of *E. coli*

under 22 different conditions allowed the validation of only 55% of the predicted open reading frames¹⁰⁰. Moreover, mass spectrometry sample preparation is time-consuming, limiting the number of experimental conditions that can be compared¹⁰¹. Alternatively, in acoustic ejection mass spectrometry, sound waves are applied to eject nanolitre droplets for analysis in a mass spectrometer, substantially increasing the sample throughput by facilitating sample preparation (that is, analysis can be performed in complex matrices without the need of extraction) for targeted proteomics and metabolomics^{102,103}. Nevertheless, targeted approaches can be used only if the compound of interest is known. By contrast, untargeted approaches can provide an overview of the total protein or metabolite content in a cell. Similar to genome annotations, untargeted metabolomics relies on spectral data from previously identified compounds to identify metabolites in a sample. However, spectral databases have only recently become available to share metabolomics data, limiting the number of entries and thus the identification of metabolites¹⁰⁴.

Proteomics data can also be combined with in vivo translation data in an approach called ribosomal profiling (Ribo-seq) to identify regulatory mechanisms at the translational level. By fixing the ribosome to the mRNA, the positions to which the ribosomes bind can be identified, indicating active translation¹⁰⁵. To fixate ribosomes at the translation initiation sites, retapamulin-assisted ribosomal profiling can be applied, which allowed the identification of more than a hundred non-conventional translation events in *E. coli*¹⁰⁶. The translation initiation site affects the translation initiation rate and thus regulates protein synthesis. However, translation can also be regulated at other stages (that is, elongation, termination or ribosome recycling), which remains poorly understood and requires further research to reveal bottlenecks that could limit the production capacity of the host.

At the metabolome level, metabolic flux analysis, or fluxomics, can aid in the identification of metabolic bottlenecks in production. Mathematical models underlying this approach provide an overview of the energy distribution throughout the cell. For example, in ¹³C-metabolic flux analysis, an isotope-labelled feedstock is used to grow bacteria and obtain an overview of the metabolites that are produced under the selected environmental conditions¹⁰⁷. Using a combination of ¹³C- and ²H-metabolic flux analyses, the periplasmic gluconate shunt in *P. putida* could be identified as an alternative pathway to glycolysis to generate pyruvate, revealing new insights into the dynamics of glucose metabolism in *P. putida*¹⁰⁸. Therefore, elucidating flux distribution can aid in rewiring the metabolism and optimizing flux towards the compound of interest.

Multi-omics integration

Although individual omics approaches can provide valuable insights into different aspects of cellular processes, their integration is crucial for obtaining a holistic understanding of bacterial metabolism. For example, the combination of transcriptomic, proteomic and metabolomic data of *P. putida* KT2440, grown on glucose in an oxygen-free anodic bio-electrochemical system, revealed that reduced acetate synthesis improves the production of 2-ketogluconate¹⁰⁹. Similarly, transcriptomics and metabolomics data of *Lactiplantibacillus plantarum* can be integrated to study its stress response during the production of linoleic acid, which has beneficial effects on human health, including the maintenance of body weight¹¹⁰.

Omics technologies can be integrated by generating a digital footprint of the bacterial metabolism to predict its performance for a certain application, for example, using genome-scale metabolic

models. Such computational models can simulate bacterial metabolism under specified environmental conditions, taking into account all biochemical reactions known for the organism of interest¹¹¹. As such, genome-scale metabolic models can predict phenotypical changes that emerge from varying environmental conditions¹¹². In addition to biochemical reactions, these models can integrate data on gene regulation (for example, transcriptomics, Ribo-seq and fluxomics) and can thus be applied not only to investigate bacterial metabolism but also to optimize production.

If the exact wiring of metabolic networks is not yet known, gap-filling algorithms can be applied to add biochemical reactions and complete the model. However, these biochemical reactions remain hypothesized and might fail to predict the organism's capacity to produce or metabolize a certain compound¹¹³. For example, the prediction by a genome-scale metabolic model that *Rhodococcus jostii* RHA1 uses ethanol as a favourable carbon source over acetate for the production of microbial lipids could not be experimentally confirmed¹¹⁴. Experimental data can aid in refining and updating genome-scale metabolic models¹¹⁵. For example, genome-scale metabolic models have been optimized for the Gram-negative non-model organism *Zymomonas mobilis* to create microbial cell factories for the optimized production of several chemicals, including succinic acid, ethylene glycol, glycolic acid, 1,4-butanediol and 1,3-propanediol¹¹⁶.

In addition, artificial intelligence, machine learning and knowledge engineering enable the integration of multi-omics data¹¹⁷. In particular, information obtained from model organisms can be combined with insights from non-model bacteria to train such algorithms. Importantly, training data need to be of high quality to build reliable models, which will require the establishment of reporting guidelines. Once optimized, these models might have a crucial role in the learning phase of the design–build–test–learn cycle¹¹⁸ by reducing turnover rates to identify patterns and engineering targets within the complex metabolic networks of non-model bacteria¹¹⁹. For example, machine learning combined with multi-omics data allowed the elucidation of the genotype–phenotype relationship of *Synechococcus* sp. PCC 7002 under different growth conditions to identify metabolic bottlenecks that can be targeted and facilitate efficient phototrophic cultivation of *Synechococcus* sp. PCC 7002¹²⁰.

Phage-based genome refactoring

The development of non-model organisms for synthetic biology requires vectors that can replicate in the non-model host and effective transformation methods to deliver the genetic material. For example, in the case of *Lactocaseilactobacilli*, electroporation remains the only effective transformation method; however, its suboptimal performance limits DNA transfer into the cell, thereby complicating the implementation of genome engineering strategies in this organism¹²¹. Moreover, reliable genetic engineering tools must be established to enable precise modifications of the host genome. Finally, well-characterized genetic building blocks (for example, promoters, terminators and transcription factors) are essential for building synthetic circuits^{14,26,122} (Table 1).

The majority of synthetic biology tools and genetic parts have been developed and optimized for model organisms, and their transfer to non-model hosts remains challenging. For example, the activity of promoters, such as *P_{em7}*, and the performance of expression systems, such as *p_{ET}* or *XylS/P_m*, can markedly vary across bacterial species^{123–125}. In addition, integrases, such as ϕ C31 and λ , rely on specific genomic attachment sites that may be absent in non-model organisms, limiting

Table 1 | Synthetic biology tools and parts for genome engineering

Function	Subtype	Applications in synthetic biology	Limitations	Strengths	Refs.
Promoters Responses vary depending on organism; orthogonality with respect to RNA polymerases	Constitutive	Synthetic (e.g., P_{am7} , P_{14a-gr} , $P_{porin/polib}$, $P_{n25-n26}$ and $P_{sl2(c)}$); native (e.g., $P_{123119r}$, P_{phaC1r} , P_{gap} and P_{pdc} , P_{eno})	Typically screened for high strength; may result in toxicity	High expression levels; does not require inducers	176–181
	Inducible	Bacterially derived (e.g., LacI/PlacUV5, XylS/Pm, AraC/PBAD, rhaRS/Prham and pSynSens)	Inducers can be expensive	Additional level of regulation; timed expression (e.g., useful for growth decoupling)	182–186
	Orthogonal	Phage-derived (e.g., pET, pPUT, pL and pSP6)	Host-specific response; additional engineering required to incorporate all elements	Function orthogonal to the host's transcriptional machinery	124,187–189
Terminators Response varies depending on organism; degradation at 3' end of transcripts renders identification of terminators difficult	Factor dependent	Rho-dependent	Cofactor is required to perform expression	Additional layer of regulation that can aid timed expression	190
	Factor independent	Phage-derived (e.g., λ : T_{10} , LUZ100: T_{19} , LUZ7: T_{50} , T7: TE); synthetic (T7 variants); bacterially derived (<i>E. coli</i> rrnB (T1))	Stability of the RNA secondary structure is affected by temperature, pH and ionic strength	Cofactor not required; based on secondary structure	135,191
Ribosome binding site	-	Synthetic (e.g., BCD1-22); native (e.g., <i>Streptomyces</i>)	Typically screened for strength; high expression may result in toxicity; experimental methods to identify ribosome binding sites remain limited	High expression levels; synthetic ribosome binding sites are designed such that they do not interfere with the secondary structure of the RNA	192,193
Aptamers	-	tBroccoli, malachite green aptamer, SP6 inhibitory aptamer, T7 inhibitory aptamer	Complex regulation; difficult to identify ligand; in vivo assays to evaluate functionality are lacking	Additional layer of regulation that can aid the development of complex circuits (e.g., biosensors)	194–197
Proteins	Integrases	ϕ C31, ϕ BT1, TG1, TP901-1, Bxb1, A118, Cre, FLP, λ , XerCD	Issues with orthogonality; integration site needs to be known and present	Development of complex synthetic circuits (e.g., memory and counting devices)	126,127,142, 198,199
	Transcription factors	Activators (e.g., ORF2 of phi31, AraC and SoxR); repressors (e.g., cl, cro, TetR and FadR), LysR	Regulation can be complex and intertwined with multiple processes, which can hinder the construction of synthetic circuits	Additional level of regulation; timed expression (e.g., useful for growth decoupling)	200–203
	Reporters	Fluorescent proteins (e.g., GFP, YFP, mCherry, RFP and BFP); luciferases (DTZ, Firefly and Renilla); chromoproteins (e.g., gfasPurple, amilCP, spisPink and eforRed)	Stability issue at high temperature or pH (e.g., thermophiles and halophiles)	Easy read-out to verify expression levels of regulatory elements enables large screens	204–208
Plasmid backbones	SEVA plasmids	pSEVA131, pSEVA237c, pSEVA2a2b8Ra	Available cargos are biased towards organisms often used for synthetic biology applications	Backbones can be used in a variety of organisms owing to the variety of available cargos	209
Genome editing tools (e.g., insertions, deletions and substitutions)	CRISPR–Cas engineering	CRISPR–Cas9, CRISPR–Cas3, CRISPR–Cas12a, CRISPR–nCas9, CRISPR–dCas9	Off-target effects; requirement of PAM sequence	Multiplex; high efficiency	128,130, 210,211
	Homologous recombination	Cre/loxP, phiC31/att, fluorescence-assisted genome engineering	Low efficiency; time-consuming; repetitive regions difficult to edit	No off-target effects	199,212
	Recombineering	λ -Red recombineering, RecT-mediated recombineering	Low efficiency; time-consuming	Short homology arms are required for recombination	213
	Transposons	Site-specific (e.g., Tn7/ <i>glmS</i>); random mutagenesis (e.g., Tn5 and Mu), CAST; OMEGA	Edits are not clean; also contain transposon sequences; specificity and editing efficiency need to be improved	High-throughput	214,215

Table 1 (continued) | Synthetic biology tools and parts for genome engineering

Function	Subtype	Applications in synthetic biology	Limitations	Strengths	Refs.
Editing gene expression levels (e.g., knockdowns)	Gene silencing	RNA interference (e.g., siRNA and miRNA); xenogeneic silencing (e.g., H-NS, Rok and CpgS)	Challenges with stability and delivery; temporary effect	Essential genes can be targeted	216,217

BFP, blue fluorescent protein; CAST, CRISPR-associated transposase; Cas, CRISPR-associated protein; dCas9, catalytically dead Cas9; DTZ, diphenylterazine; *E. coli*, *Escherichia coli*; GFP, green fluorescent protein; mCherry, monomeric red fluorescent protein; miRNA, microRNA; nCas9, Cas9 nickase; OMEGA, Obligate Mobile Element Guided Activity; ORF2, open reading frame 2; RFP, red fluorescent protein; SEVA plasmids, Standard European Vector Architecture plasmids; siRNA, small interfering RNA; YFP, yellow fluorescent protein.

their utility^{126,127}. Alternatively, analogous tools and parts can be directly isolated from the genome of non-model species or their phages, serving as blueprints for the rational design of customized synthetic biology toolkits. For example, owing to its high tolerance for chemicals and solvents, *P. putida* has been explored as a chassis for biotechnological applications⁷⁵, including the development of genome engineering tools, such as fluorescence-assisted genome engineering, CRISPR–Cas3 engineering as well as a cytidine base editor^{128–130}. Furthermore, regulatory elements derived from bacteriophage phi15 have been used to develop a *Pseudomonas*-specific counterpart of the pET system, enabling tightly controlled recombinant protein expression within this genus¹²⁴.

Bacteriophages as sources of regulatory elements

Bacteriophages provide a rich source of molecular biology tools and circuits not only for model organisms, such as *E. coli*, but also for non-model microbes (Table 1). These viruses specifically infect bacteria, hijacking their metabolism to propagate and form new virus particles¹³¹. Bacteriophages are equipped with a variety of genetic building blocks (that is, promoters, terminators, integrases and repressors) that function independently from their bacterial host and that are evolutionarily optimized to work well within that host. Therefore, their genomes can be mined to design a toolbox for non-model organisms that functions orthogonally to their native metabolism^{132,133}.

Phages are also valuable for the discovery of regulatory elements, for example, for the identification and functional characterization of phage-derived promoters and terminators (Fig. 3). Here, a transcriptomics approach (that is, ONT-cappable-seq) can be used to precisely locate regulatory elements in the phage genome^{95,134}. A set of fluorescence-based in vivo assays can then be applied to validate the strength and functionality of these regulatory elements in the host of interest^{135,136}. In the case of phage-derived promoters, a distinction needs to be made between host-specific phage promoters (that is, those recognized by the host RNA polymerase) and phage-specific phage promoters, whose expression requires the presence of a phage-encoded RNA polymerase. Cloning the promoter upstream of a fluorescent reporter (for example, GFP) can aid in assessing its strength¹³⁶ (Fig. 3). By contrast, for phage-derived terminators, the termination activity, which is a measure for the amount of read-through at the termination site, can be determined by cloning the phage-derived terminator in between two fluorescent proteins¹³⁵ (Fig. 3). As such, the functionality of genetic building blocks derived from phages can be screened in a rapid, high-throughput manner. Using this approach, a large set of phage promoters and terminators have been identified and their functionality tested in vivo in different *Pseudomonas* species^{135,137,138}. Phages can also encode their own transcriptional machinery or sigma-like proteins that redirect the host RNA polymerase to favour phage-derived sequences, which can be used to

further fine-tune transcriptional regulation^{136,139,140}. Although efforts have mainly focused on transcriptional regulation thus far, phage genomes may also encode translational elements, such as ribosome binding sites and riboswitches, which could be exploited for synthetic biology in non-model hosts¹³².

Tools for genome engineering and posttranslational modifications

Phages are also equipped with tools to facilitate genome engineering. For example, temperate phages naturally integrate into the bacterial chromosome using integrases. These enzymes, in particular, phage λ-derived red recombinase, can be exploited to provide access to the genome of non-model bacteria¹⁴¹ (Table 1). Integrases can also be further engineered or additional ones can be mined from phages to provide accessibility to a larger number of genomes^{142,143}.

In addition to regulation at the DNA level, posttranslational modifications can fine-tune bacterial metabolism¹⁴⁴. Posttranslational modifications, such as RNylation, phosphorylation and acetylation, also occur in phages to regulate bacterial responses related to virus–bacteria interactions as well as bacterial transcription and translation¹⁴⁵. Insights into the interplay between phages and their hosts could aid in improving phage resistance in bacterial fermentations, as phage contamination remains a risk in bacterial fermentations and solutions remain limited¹⁴⁶. Finally, the ability of phages to kill their bacterial host can be used for engineering purposes. As genetically engineered organisms should not be released into the environment, microbial kill switches are necessary to contain modified organisms¹⁴⁷. Phage proteins that are involved in lysing bacteria could be converted into synthetic circuits to improve biosafety¹⁴⁸.

Industrial applications

Endogenous gene expression in native hosts can be applied in biotechnology, including pharmaceutical, agricultural, fresh water, marine and desert biotechnology¹⁴⁹, in particular, in scenarios where heterologous gene expression underperforms owing to physiological constraints (Table 2).

Pharmaceutical industry

Non-model organisms can be used as microbial cell factories, for example, for the identification of new antimicrobials and anticancer agents^{150–152}. Owing to their underexplored genomes, non-model organisms offer a valuable source of new bioactive compounds with pharmaceutical potential. However, the complexity of many biosynthetic pathways makes heterologous expression of the gene clusters encoding these compounds challenging. Native hosts are naturally optimized and have evolved innate resistance mechanisms to tolerate high quantities of many of these compounds and are thus suitable for antimicrobial and anticancer agent production (Table 2).

Chemical and agricultural industry

Endogenous expression in non-model organisms can also be applied in the chemical and agricultural industry (Table 2). For example, the methanotrophic bacteria *Methylococcus capsulatus* can be used to produce a protein-rich feed, which can be applied in agriculture, the pet industry and aquacultures¹⁵³. In addition, lactic acid can be produced in *Lactobacilli* for use in bioplastics or as agrochemical in crop production. Moreover, entire bacteria, such as *Bacillus*, *Pediococcus* and *Lactobacillus*, can be applied in wastewater treatment¹⁵⁴. Bioherbicidal compounds, such as coronatine, can also be produced in microorganisms. However, the production titres in the native host *Pseudomonas syringae* pv. *tomato* DC3000 are low. The coronatine biosynthetic gene cluster can also be heterologously expressed in *P. putida* KT2440, albeit with similarly low production titres. Moreover, heterologous production leads to an imbalance between the coronafacic and coronamic acid moieties, which together form coronatine. The engineered *P. putida* strain produces an excess of coronafacic acid relative to coronamic acid, which compromises the eventual titre of coronatine¹⁵⁵.

Furthermore, pigments for the textile and paint industry can be produced using colourants derived from *Streptomyces coelicolor*¹⁵⁶.

In addition, prodigiosin is a red pigment produced by *Serratia marcescens* at titres of up to 36 g l⁻¹ (ref. 157). However, *S. marcescens* is pathogenic, and the regulatory mechanisms that drive prodigiosin production in this species are complex and intertwined with many metabolic processes, which complicates heterologous expression^{158,159}. Alternatively, prodigiosin can be heterologously produced in *P. putida* KT2440, albeit at a low production titre of 0.094 g l⁻¹, probably owing to a stress response¹⁶⁰.

Oleaginous non-model bacteria can be valuable as palm oil substitutes and oil sources^{117,161,162}. In addition, *Rhodococcus opacus* might serve as host for biodiesel production, because it can accumulate lipids of up to 70% of its dry weight. These lipids can be used as feedstocks for the production of biodiesel owing to their high calorific value¹⁶³. Biosurfactants, which are commonly applied as detergents or emulsifiers, can also be synthesized by non-model organisms. For example, *Burkholderia* and *Pseudomonas* strains are natural producers of rhamnolipids, a class of biosurfactants. These native strains can achieve production titres up to 2.5 times higher than those observed in heterologous expression systems using model organisms^{164,165}. Owing to the antimicrobial properties of rhamnolipids, it is hypothesized that native

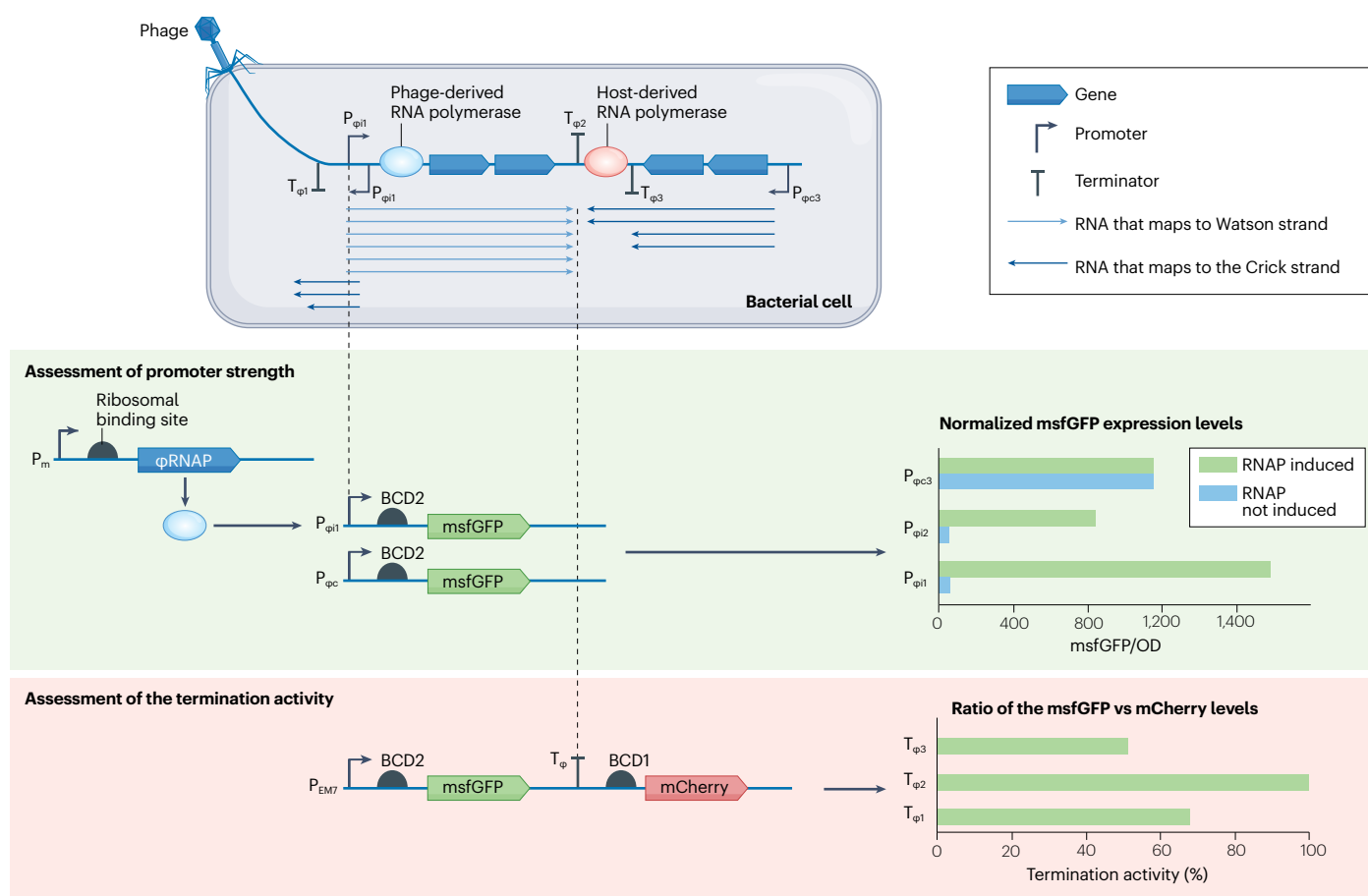


Fig. 3 | Pipeline to identify and test phage-derived regulatory elements. Transcriptomics is performed to identify promoter and terminator regions. Subsequently, their functionality can be tested in fluorescence-based assays to assess their functionality in the host of interest. In the case of promoters, the level of fluorescence intensity is assessed, whereas in the case of terminators, the termination activity is calculated. The latter is a measure of the fluorescence

intensity of two subsequent fluorescent proteins, where a construct with a terminator is compared to a vector without terminator¹³⁵. BCD, bicistronic design; GFP, green fluorescent protein; OD, optical density; P_{EM7}, constitutive promoter; P_m, inducible promoter; P_{ph1}, host-specific phage promoter; P_{ph2}, phage-specific phage promoter; RNAP, RNA polymerase; T_{ph}, phage terminator.

Table 2 | Industrial applications of non-model bacteria and their metabolites

Industrial branch	Application	Compound/ enzyme	Microorganism	Examples of products ^a
Pharmaceutical industry	Antibiotic	Kanamycin	<i>Streptomyces kanamyceticus</i>	Kantrex
		Tetracycline	<i>Streptomyces aureofaciens</i>	Tetracycl, Panmycin
		Chloramphenicol	<i>Streptomyces venezuela</i>	Chloromycetin Ophthalmic, Chloroptic
		Erythromycin	<i>Saccharopolyspora erythraea</i>	Erythrocin, Erythroped, Erymax
		Vancomycin	<i>Amycolatopsis orientalis</i>	Edicin
		Clindamycin	<i>Streptomyces lincolnensis</i>	Duac Acne, Treclinac
		Rifamycin	<i>Amycolatopsis rifamycinica</i>	Aemcolo
		Cycloserine	<i>Streptomyces orchidaceus</i>	Seromycin
		Pristinamycin	<i>Streptomyces pristinaespiralis</i>	Pyostacine
		Fosfomycin	<i>Streptomyces fradiae</i>	Monurol, Monuril, Ivozfo
		Meropenem	<i>Streptomyces cattleya</i>	Merrem
		Daptomycin	<i>Streptomyces roseosporus</i>	Cubicin, Dapzura RT
		Fidaxomicin	<i>Dactylosporangium aurantiacum</i> ssp. <i>hmdensis</i>	Dificlir
		Gramicidin	<i>Bacillus brevis</i>	Neosporin
		Bacitracin	<i>Bacillus subtilis</i>	BACiiM
		Colistin	<i>Paenibacillus polymyxa</i>	Xylistin, Coly-Mycin M, Colobreathe
		Mupirocin	<i>Pseudomonas fluorescens</i>	Bactroban, Centany
		Aztreonam	<i>Chromobacterium violaceum</i>	Azaxtam
	Anticancer agent	Bleomycin	<i>Streptomyces verticillus</i>	Blenoxane
		Dactinomycin	<i>Streptomyces parvullus</i>	Cosmegen
		Mitomycin C	<i>Streptomyces caespitosus</i>	Jelmyto, Mitosol
		Doxorubicin	<i>Streptomyces peucetius</i> var. <i>caesius</i>	Adriablastina, Caelyx, Doxorubin
		Daunorubicin	<i>Streptomyces peucetius</i>	Cerubidine
		Epirubicin	<i>Streptomyces peucetius</i>	Ellence
	Antifungal	Clotrimazole	<i>Streptomyces antibioticus</i>	Canesten
		Nystatin	<i>Streptomyces noursei</i>	Mycostatin, Nyamyc
		Amphotericin	<i>Streptomyces nodosus</i>	Fungizone, Mysteclin-F, AmBisome
	Anthelmintic	Ivermectin	<i>Streptomyces avermitilis</i>	Stromectol
	Probiotics	–	<i>Limosilactobacillus reuteri</i>	Pylopass
		–	<i>Lactobacilli</i>	ACN, VGN, FNG
Chemical industry	Pigments	Anthranilic acid	<i>Streptomyces coelicolor</i>	Pili biobased anthranolic acid
	Cleaning agents	–	Combination of 5 bacillus strains	Microvia
	Antiaging compound	Phytoene	<i>Deinococcus geothermalis</i>	Phyt-N-Resist
Agricultural industry	Biologicals	–	Combination of <i>Penicillium bilaiae</i> , <i>Bacillus amyloliquefaciens</i> and <i>Trichoderma virens</i>	BioniQ
		–	<i>Bradyrhizobium arachis</i>	Cell-Tech
		–	<i>Bacillus amyloliquefaciens</i>	Taegro 2
		–	<i>Bacillus thuringiensis</i>	VectoBak, Dipel, FlorBac
		–	<i>Bacillus amyloliquefaciens</i>	Toreda
		–	Consortium of aerobic bacteria in combination with <i>Rhodobacter</i> sp.	AMF

Table 2 (continued) | Industrial applications of non-model bacteria and their metabolites

Industrial branch	Application	Compound/ enzyme	Microorganism	Examples of products ^a
Agricultural industry (continued)	Probiotics	–	<i>Bacillus subtilis</i>	Alterion
	Protein-rich feed	–	<i>Methylococcus capsulatus</i>	Uniprotein
	Agrochemicals	Abamectin	<i>Streptomyces avermitilis</i>	Apache, Vivid II
		Spinosad	<i>Saccharopolyspora spinosa</i>	Tracer, Conserve
		Bilanafos	<i>Streptomyces hygroscopicus</i>	Bialaphos
	Antifouling agent	Entire secretome	<i>Pseudomonas</i> strain PF-11	BMX-11
Pet industry	Anthelmintic	Milbemycin oxime	<i>Streptomyces hygroscopicus aureolacrimosus</i>	Interceptor
	Ammonia removal in ponds	–	Nitrifying bacteria	Bactocool
Food industry	Low-lactose dairy products	Lactase	<i>Kluyveromyces lactis</i>	Lactozym, HA-Lactase, Maxilact
	Fish substitutes	Cellulose	<i>Komagataeibacter</i> , <i>Acetobacter</i> , and <i>Gluconacetobacter</i>	Tuna, Scallops
	Pulque (agave wine)	–	<i>Zymomonas mobilis</i>	The Queen La Reina Pulque
	Probiotics	–	<i>Lactocaseibacillus casei</i> , <i>Lactobacillus acidophilus</i> , <i>Lactiplantibacillus plantarum</i> , <i>Bifidobacterium bifidum</i>	Activia probiotic yoghurt
		–	<i>Lactobacillus acidophilus</i> , <i>Bifidobacterium animalis</i> subsp. <i>lactis</i> and <i>Lactocaseibacillus rhamnosus</i>	Optimel Probiotica

^aProduct names are limited to three examples.

producers have evolved enhanced tolerance to these compounds, offering a potential advantage over heterologous hosts.

Industrial waste streams

Industrial waste streams might be reduced by adopting circular processes to recycle and reuse waste. In particular, industrial waste streams that contain nutrients can serve as feedstocks for bacteria to generate valuable new products. For example, lignin, with an annual global production of 34 billion tons¹⁶⁶, can be degraded by a variety of bacterial species, including *P. putida*, *E. coli* and *Corynebacterium glutamicum*. These microorganisms are capable of converting lignin into high-value compounds, such as vanillin or coumaric acid¹⁶⁶. Crude glycerol and fruit peel waste streams can also be upcycled through microbial fermentation. These materials are rich in carbohydrates and, thus,

require minimal supplementation to support bacterial metabolism¹⁶⁷, thereby reducing costs. However, the exploration of waste streams as feedstocks remains limited owing to the metabolic constraints of model organisms, which often lack the metabolic capacities to grow on these substrates¹⁶⁷.

Outlook

Endogenous gene expression in non-model organisms, which have been evolutionary optimized to produce certain proteins or metabolites in response to environmental stimuli, might offer an alternative route in synthetic biology^{26,75,168}. In particular, multi-omics techniques enable the identification of bacteria for specific synthetic biology and biotechnological applications¹⁶⁸ (Box 1). The use of non-model organisms, however, requires a shift to fundamental rather than

Box 1 | High-throughput versus rational-based approaches

The microbial synthetic biology market is expanding. With a compound annual growth rate of 25%, the global market is projected to be worth US\$148.93 billion by 2033²²³. For example, Ginkgo Bioworks has developed a service platform to engineer microbial strains for the production of compounds of interest. With the implementation of high-throughput automated strategies, a multitude of strains can be screened to optimize production. This platform runs on a limited set of microorganisms, which are trained to produce proteins or metabolites by adaptive laboratory evolution²²⁴. During this process, bacteria are exposed to specific environmental conditions to improve their survival and production rates under the selected conditions. However, adaptive laboratory evolution remains unpredictable and might cause mutations that do not favour the application of interest. Therefore, non-selective, high-throughput methods have to be applied to screen a large

number of strains²²⁵, requiring a large amount of resources. Furthermore, the number of generations required to reach an improved production strain are unpredictable, which makes this approach time-consuming and cost-prohibitive²²⁶.

Alternatively, a deliberate selection of chassis can be tested for a given application to reduce the costs and time required to optimize microbial fermentations. This rational-based approach can aid in making targeted alterations in the bacterial genome to optimize production. Instead of bypassing evolution, fine-tuned, inherent processes can be exploited. For example, Geno(matica) applies this approach to improve production titres and for sustainable microbial fermentation²²⁶. Production costs remain a key hurdle in the marketing of microbial cell factories. Therefore, rational-based engineering approaches based on endogenous expression should be considered in microbial synthetic biology.

application-oriented research to improve our understanding of non-model bacteria and develop synthetic biology tools to access their genomes. In part, phages can offer a solution, as their genomes are equipped with genetic building blocks that are optimized to function in their bacterial hosts¹³².

Synthetic biology has largely concentrated on replacing traditional petrochemical processes. As a result, bacteria have been engineered as microbial cell factories to produce a range of chemicals for use in agriculture, medicine and industrial biotechnology. However, even with optimized fermentation strategies, microbial production titres often fall short of those achieved through conventional chemical synthesis, limiting their competitiveness in certain applications. Accordingly, scale-up remains a key technical bottleneck to be addressed. In addition to high operational costs and expensive growth media, culture conditions substantially differ upon scale up, which should be taken into account in the development of a production strain¹¹⁷. To facilitate scale-up, bioreactors can be equipped with sensors (for example, optical and electrical sensors) to monitor the production process¹⁶⁹. Furthermore, environmental bioreactor conditions can be mimicked by laboratory-scale equipment to optimize microbial cell factories as industrial-scale fermentation hosts¹⁷⁰. Downstream processing operations can also vary in complexity and add to the production costs¹⁷¹. Moreover, the genetic stability of the strain needs to be considered, as bacteria can have spontaneous mutation rates of 10^{-9} – 10^{-10} per nucleotide per generation¹⁷². As growth in bioreactors induces stress, these mutation rates can further increase. Therefore, the genetic stability of the strain should be traced over time to ensure steady production levels.

The application of non-model organisms in the field requires thorough knowledge of their biology. Under the current regulatory framework in Europe, only products derived from genetically modified microorganisms (GMMs) in contained use, such as in industrial fermentation, are permitted for commercialization. By contrast, the release and marketing of GMMs, particularly in healthcare and agriculture, are subject to much stricter regulations and are often restricted or prohibited¹⁷³. A key concern is safety with regard to human health and the environment. Therefore, risk assessments are in place to assess their safety concerning toxicity, horizontal gene transfer and traceability¹⁷⁴. This intensive risk assessment process, in combination with the lack of centralized regulation both in the EU and globally, might hamper the investment in technologies that explore GMMs^{173,175}. Nevertheless, GMMs are being explored for a variety of applications, including live-attenuated vaccines, probiotics and bacterial batteries^{10–12}.

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J.P. researched data for the article and contributed to discussion of content and writing. M.B., M.D.M., J.M. and R.L. contributed to discussion and reviewing/editing the manuscript before submission.

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