



# Metabolomics tools for the synthetic biology of natural products

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# Metabolomics tools for the synthetic biology of natural products

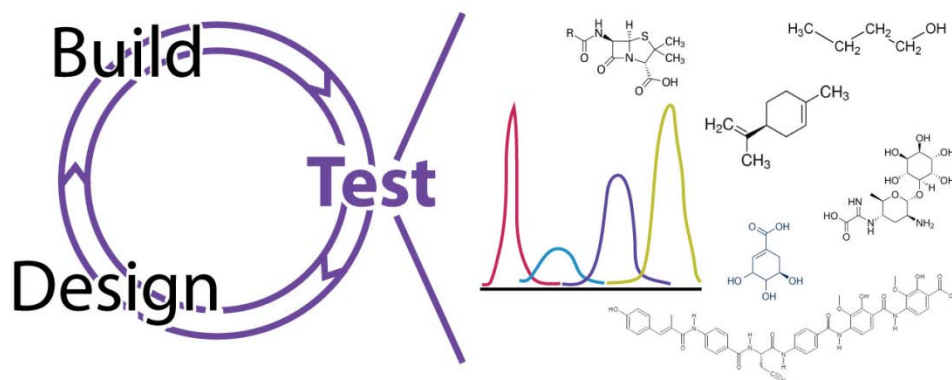
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## Abstract

Metabolomics plays an increasingly central role in within the Design – Build – Test cycle of synthetic biology, in particular in applications targeting the discovery, diversification and optimized production of a wide range of natural products. For example, improved methods for the online monitoring of chemical reactions accelerate data generation to be compatible with the rapid iterations and increasing library sizes of automated synthetic biology pipelines. Combinations of label-free metabolic profiling and  $^{13}\text{C}$ -based flux analysis lead to increased resolution in the identification of metabolic bottlenecks affecting product yield in engineered microbes. And molecular networking strategies drastically increase our ability to identify and characterize novel chemically complex biomolecules of interest in a diverse range of samples.

## Graphical abstract



## Synthetic biology of natural products

Synthetic biology facilitates the biosynthesis of pharmaceutical ingredients and other high-value chemicals by employing the Design – Build – Test cycle of engineering to guide the systematic enhancement of microbial factories [1-4]. Exemplary successful applications of synthetic biology to natural product production include a one-pot method for menthol biosynthesis in *Escherichia coli* [5], the modular extension of a styrene biosynthesis pathway to produce 2-phenylethanol [6], cannabinoid biosynthesis in yeasts [7], and the heterologous production of antibiotics using extensively refactored biosynthetic gene clusters: myxobacterial  $\alpha$ -pyrone antibiotics in *Myxococcus xanthus* [8] and kasugamycin (an aminoglycoside antibiotic isolated from *Streptomyces kasugaensis*) in actinomycetes [9]].

A recent review by Smanski and colleagues [10] provides details of recent advances in the technologies underpinning the Build aspects of the synthetic biology cycle, including pathway construction and pathway screening, while a complementary review by Chen *et al.* [11] focuses

43 on the modelling approaches for the construction and optimisation of cell factories for bio-  
44 production, which cover a large part of the Design activities. In the present review, we will  
45 turn focus on the Test component of synthetic biology, focusing in particular on advances in  
46 metabolomics as a discovery and debugging tool for metabolically enhanced microbial systems.

47  
48

## 49 **Test Analytics – Appropriate Technologies**

50

51 Mass spectrometry (MS) coupled to chromatography remains the domineering technology used  
52 for the quantification of natural product targets and is also the most widely used platform for  
53 the global profiling of the impact of an engineered biosynthetic pathway on the microbial  
54 metabolome. The challenge for the analytical technologies is to achieve the acquisition speed  
55 and sensitivity required to meet the high-throughput needs of a synthetic biology-based  
56 pipeline. Traditionally, products are measured directly from an aliquot of cell culture medium  
57 or – in the case of volatile products – they are captured in solvent overlays and transferred to  
58 vials or multi-well plates for analysis. These approaches are often slow (tens of minutes per  
59 sample plus preparation time) and provide only a snap shot of what is occurring at a given time.

60

61 To overcome this limitation, much effort has been invested into the development of improved  
62 methods for the online monitoring of chemical reactions; which would provide greater control  
63 of sampling and provide dynamic results with regards to product turnover. **Definitions of the**  
64 **analytical terminology described herein are summarised in Table 1.** In recent work by Yan *et al.*,  
65 desorption electrospray ionization (DESI) coupled to ion-mobility mass spectrometry was used  
66 for the high-throughput screening of biocatalysis directly from bacterial colonies on agar plates  
67 [12], which can in principle be applied to a broad range of substrates and products, including  
68 free amines, carboxylic acids, alkaloids and phenols; multiple analytes can be detected in a  
69 single analysis thus allowing for the screening of diverse strain libraries with complex product  
70 profiles. DESI-MS was also applied to the rapid analysis of enzyme kinetics by Cheng and co-  
71 workers [13], who measured product formation in a buffered aqueous medium, explored the  
72 possibility of adjusting the pH and solvent composition of the DESI spray to quench the  
73 enzymatic reaction and thus improved the accuracy of the kinetic measurements by preventing  
74 post-ionization reactions.

75

76 As an alternative to DESI, matrix-assisted laser desorption ionisation mass spectrometry  
77 imaging (MALDI-MSI) has readily been applied towards the large-scale phenotyping of bacteria  
78 [14, 15]. A related optically-guided MALDI-MS strategy has recently been implemented for the  
79 profiling of microbial colonies for rapid screening of natural product analogue libraries [16].  
80 This impressive development used optical imaging of microbial colonies to direct the laser  
81 coordinates for an automated MALDI-MS screening of approximately 1000 colonies directly  
82 from an imprinted glass slide with an MS sampling rate of about one colony per second.  
83 Reaction products were screened *in situ* and results overlaid with the optical images;  
84 integration of results allowed for subsequent colony picking and recovery of the desired mutant  
85 strains. The majority of commercially available MALDI-MS instrumentation permit a spatial  
86 resolution of > 100 µm. However, the group of Bernhard Spengler has recently dramatically  
87 pushed this boundary towards much better lateral resolutions down to 1.4 µm [17], thus further  
88 advancing the technique towards single cell resolution and even higher throughput [18].

89

90 The coupling of microreactors or continuous flow chemical reactors directly to the mass  
91 spectrometer provides an enhanced ability to characterise unstable reaction products and  
92 reduces the sample volume required (albeit with sufficient mass spectrometer sensitivity). **Link**  
93 **et al. [19] provided a comprehensive example of such an application. They**  
94 **demonstrated the ability to undertake real-time metabolome profiling by direct**

95 injection of living bacteria, yeast and mammalian cells into a high-resolution mass  
96 spectrometer through coupling a peristaltic pump and two six-port valves and  
97 automatically sampling from a liquid culture. This approach permitted the automated  
98 monitoring of around 300 compounds in 15–30 s cycles over several hours. They investigated  
99 the metabolite dynamics in real-time during 2 h starvation and 30 min of growth resumption.  
100 The approach suggested that the accumulation of energetically costly metabolites in starved *E.*  
101 *coli* reflects the control strategy to favour cheap metabolic pathways for growth resumption.  
102 From an analytical perspective the method permitted real-time metabolome profiling that  
103 followed the dynamics of metabolic processes in different organisms over extended periods.  
104 The method alleviates retrospective manual sampling, sample preparation and sample  
105 manipulation associated with traditional off-line methods.

106  
107 Progress has also been made on the mass spectrometry techniques available to the synthetic  
108 biology community: proton transfer reaction mass spectrometry (PTR-MS) and selected ion  
109 flow tube mass spectrometry (SIFT-MS). These techniques are direct injection approaches that  
110 utilise chemical ionisation for real time analysis of volatile organic compounds. PTR-MS has  
111 been shown to achieve near-to-real-time monoterpene separation and identification, when  
112 coupled to a fast gas chromatography, with sensitivity in the range of 1.2 ppbv from plant  
113 material [20]. PTR-MS has also been applied to the real time monitoring of the yeast volatilome  
114 [21], detecting more than 300 metabolite features, 70 of which were tentatively identified, in  
115 the headspace of *Saccharomyces cerevisiae* cultures over 11 days at 4-h time points. Additional  
116 development and application of this technique has been demonstrated by Materic *et al.* [22],  
117 who used Selective Reagent Ion PTR-MS to investigate the separation of monoterpene mixtures,  
118 which are a particularly common target in recent synthetic biology projects *i.e.* geraniol [23],  
119 linalool [24] and limonene [25].

## 122 **Global analysis – Metabolomics**

123 Synthetic biology requires not only the rapid and accurate quantitation of the desired end  
124 products; even more important for a systematic engineering of the microbial factories is a  
125 thorough understanding of metabolic flux and the regulation of central carbon metabolism to  
126 ensure the desired production of target compounds is compatible with maintaining cellular  
127 homeostasis and energy balance. Metabolomics, the comprehensive profiling of small molecules  
128 in a biological sample, is the obvious method of choice for collecting the necessary data for this  
129 kind of analysis, and synthetic biology can build on a continuously refined repertoire of  
130 metabolomics approaches [26, 27].

131  
132 Of the many technological advances in recent years, we only highlight the increasing importance  
133 of parallel reaction monitoring (PRM) in metabolomics; the quantitation of intermediates of  
134 central carbon metabolism, amino acids and shikimate pathway-related metabolites in  
135 engineered strains of *E. coli* [28] is just one important example of its application in synthetic  
136 biology. PRM permits the quantitative analysis of multiple targets (237 in this example) [29]  
137 with excellent linearity of quantitation, as well as high precision and accuracy. In a related  
138 approach, all ion fragmentation acquisition has recently been demonstrated to achieve  
139 increased accuracy in metabolite identification for a large number of pre-selected compounds,  
140 while at the same time acquiring full scan information to allow the identification of additional  
141 metabolites that were initially not targeted [30].

142  
143 A metabolomics-driven approach was applied to identify non-obvious target genes to further  
144 improve the production of 1-butanol [31, 32]. The authors performed quantitative targeted  
145 analysis of acyl-CoAs in the CoA-dependent 1-butanol biosynthetic pathway in *Synechococcus*  
146 *elongatus* strains *via* <sup>13</sup>C-labelling of cell extracts as an internal standard and HPLC-MS analysis.  
147 The results indicated several targets for potential improvements of 1-butanol production in

148 cyanobacteria, such as possible rate-limiting steps (reductive reaction of butanoyl-CoA to  
149 butanal) or effective regeneration of free-CoA from butanoyl-CoA to enhance the conversion of  
150 pyruvate to acetyl-CoA. In a parallel study addressing 1-butanol production in *E. coli*, the  
151 authors examined the metabolomic impact of the deletion of phosphate acetyltransferase, which  
152 was performed in an attempt to reduce the amount of acetate produced and simultaneously  
153 increase the acetyl-CoA pool. Metabolomics analysis using a targeted ion pair LC-MS/MS  
154 method detected a total of 78 metabolites and pointed to several metabolic perturbations  
155 caused by the deletion that seemed to be the consequence of a CoA imbalance or insufficient  
156 CoA recycling, which caused the undesirable accumulation of side products. Further  
157 metabolomics analysis identified the underlying enzymatic bottleneck, alcohol dehydrogenase,  
158 and fine-tuning of this activity resolved the CoA imbalance and led to substantially improved 1-  
159 butanol titres [31].

160  
161 A metabolomics approach was also implemented to investigate central metabolism of a fructose  
162 repressor (*fruR*) knockout in a recombinant L-tryptophan producing strain of *E. coli* (*E. coli* FB-  
163 04) [33]. The authors report more than 80 intracellular metabolites that were altered as a result  
164 of the knockout, 23 of which were related to tryptophan biosynthesis. The levels of glycolysis,  
165 pentose phosphate and TCA cycle intermediates were consistently increased, and levels of  
166 shikimate derivatives (direct tryptophan precursors) and L-glutamine were decreased in the  
167 knockout strain, which also showed a substantially increased tryptophan production. The  
168 interpretation of these results illustrates very clearly the pitfalls of using steady-state  
169 metabolome profile information as a proxy for metabolic fluxes, which are of central interest for  
170 synthetic biology: based on increased levels of glycolytic and pentose phosphate pathway  
171 intermediates, the authors conclude that the *fruR* knockout enhanced metabolic flow through  
172 these two pathways which provide the substrates for L-tryptophan biosynthesis. However, the  
173 TCA cycle, which directly competes with tryptophan biosynthesis shows an equally increased  
174 level of its intermediates, and the only pathway for which direct flux measurements are  
175 available, tryptophan biosynthesis itself, shows a consistent decrease in its key intermediates,  
176 despite an increase in flux by 62.5% (from 0.024 to 0.039 g/L/h).

177  
178 A subsequent study combining metabolomics and <sup>13</sup>C fluxomics provided more detailed insights  
179 into the metabolic flux redistribution in an *E. coli* strain overproducing shikimic acid with high  
180 titres and yields: Rodriguez *et al.* [34] used an engineered AR36 *E. coli* strain constitutively  
181 expressing six proteins encoded in a synthetic operon promoting high-yield production of  
182 shikimic acid from glucose. Comparative metabolomics of a production strain and parental  
183 strains (carrying either no plasmid or “empty plasmid”) was used to track the levels of seven  
184 exometabolites and 25 endometabolites over time. It revealed a global remodelling of carbon  
185 and energy metabolism in the high producer. This resulted in reduced carbon available for  
186 oxidative and fermentative pathways and increased levels of endometabolites involved in  
187 energy pathways, preventing the depletion of essential intermediates, such as PEP and ATP.  
188 Both glycolytic flux and TCA cycle activity were substantially reduced in this overproduction  
189 scenario (43 g/L of shikimate in 30 h on complex medium).

190  
191 Given its importance as a provider of essential precursors for a diverse range of  
192 biotechnologically important biochemicals, it is not surprising that the shikimate pathway has  
193 been the target of dedicated metabolomics method development: *e.g.*, Lai *et al.* [35] contributed  
194 a robust HPLC method for the quantification of aromatic substrates, products and pathway  
195 intermediates in order to accelerate strain engineering for industrial production of aromatics as  
196 biosynthetic molecules. The achieved limits of detection between 10<sup>-10</sup> – 10<sup>-13</sup> mol make the  
197 method suitable for endometabolome and exometabolome analysis of engineered strains.

198  
199 Another example of a metabolomics-based strategy for strain engineering (this time utilising a  
200 GC-MS analytic platform) is the study by Teoh *et al.* [36] investigating phenotypic differences in

201 growth rates and metabolite profiles of nineteen single-deletion *S. cerevisiae* mutant strains  
202 cultivated under stress-free and under 1-butanol stress conditions (growth inhibition caused by  
203 higher alcohols (*e.g.* 1-butanol) is considered as a bottleneck in their biosynthetic production).  
204 Metabolites associated with improved growth rates under stress conditions were identified, and  
205 new stress-resistant mutant yeast strains were successfully predicted based on their metabolite  
206 profiles. This approach illustrates the potential of metabolomics as a predictive screening tool to  
207 inform semi-rational strain engineering approaches.

208  
209 Finally, metabolomics has been applied for the monitoring of isoprenoid precursors production,  
210 another classic target for synthetic biology [37, 38]. In a study by Kirby *et al.* [39], who report  
211 for the first time the functional expression of an extensively engineered functional 1-deoxy-D-  
212 xylulose 5-phosphate (DXP) pathway in *S. cerevisiae* which normally utilizes the mevalonate  
213 pathway, which has a lower theoretical yield. Metabolite-guided DXP pathway balancing, by LC-  
214 MS quantification of intermediates in cultures exhibiting various levels of flux, appeared to be a  
215 successful approach for identifying a bottleneck in the pathway. An engineered strain  
216 exclusively using the DXP pathway achieved an endpoint biomass 80% of that of the same strain  
217 using the mevalonate pathway under low aeration conditions.

218  
219

## 220 **Molecular networking – moving forward**

221 The main challenge of untargeted metabolomics is compound annotation; the persistent  
222 difficulties of confidently identifying the detected metabolites currently seriously limits the  
223 utility of the MS data acquired. Molecular networking, a visualisation method for tandem MS  
224 data, is a powerful complement to traditional de-replication methods [40]. This approach allows  
225 for the detection of sets of spectra from related molecules (“spectral networks”), even in the  
226 cases when these spectra are not matched to any known compounds. The approach is based on  
227 the assumption that similar molecules have similar MS fragmentation patterns so they will tend  
228 to cluster closely within a network. Each spectrum (ideally derived from a single compound) is  
229 visualised as a network node, and the edges between nodes represent a degree of similarity  
230 between spectra. The thicker the line, the more MS/MS fragment ions are shared by the two  
231 connected nodes. Nodes can be supplemented by such information as a compounds abundance,  
232 biochemical activity, origin *etc.* Molecular networking led to the development of Global Natural  
233 Products Social Molecular Networking (GNPS), a metabolomic data-driven platform for the  
234 storage, sharing, analysis, and knowledge dissemination of tandem MS spectra where one is  
235 able to annotate natural product data *via* continuous de-replication. [41].

236  
237 Although improvements are still required to obtain unambiguous analysis of molecular  
238 networks such as efficient integration with existing LC-MS detection strategies, enhancement of  
239 pre-processing and universal optimal acquisition methods [40, 42, 43], its applications are  
240 expanding fast [44-49] and it will soon become an indispensable metabolomics tool in  
241 exploratory analyses for the synthetic biology of novel natural products. For example, in a  
242 recent study Crüseman and colleagues [46] screened 146 marine *Salinispora* and *Streptomyces*  
243 strains using HPLC-MS/MS, molecular networking, and the Global Natural Products Social  
244 (GNPS)[41] platform and explored the impact of differing culturing and extraction techniques.  
245 The systematic investigation of the effect of these parameters clearly demonstrated how much  
246 inherent chemical diversity could be missed when just one culture and extraction protocol is  
247 utilised to assess metabolic capacity. This example demonstrated how the application of  
248 molecular networking permits the rapid optimisation of experimental parameters that can  
249 subsequently be implemented early in the discovery workflow.

250  
251 Okada *et al.* [50] used molecular networking for the investigation of the influence of  
252 trimethoprim (Tmp) antibiotic on the secreted metabolome of *Burkholderia thailandensis* E264.  
253 The untargeted comparison of Tmp-induced and uninduced samples (utilising HPLC-QToF-

254 MS/MS) resulted in ~240 metabolites of interest (with >100 compounds observed only for the  
255 induced samples). Organising them into 14 sub-networks followed by NMR analysis enabled  
256 rapid identification of 40 compounds including analogues of known compounds and a group of  
257 new molecules, acybolins, showing that molecular networking aids rapid identification of  
258 compounds compared to traditional workflows.

259  
260 In related work, von Eckardstein *et al.* [47] used bioactivity-guided untargeted LC-MS/MS  
261 analysis and molecular networking in the search of new antibiotic agents from *Xanthomonas*  
262 *albilineans*. Over 20,000 MS/MS spectra acquired from crude extracts and bioactive fractions  
263 were organised into a molecular network *via* the GNPS portal, which allowed for the  
264 identification of potential derivatives in the albicidin sub-network. The group reported eight  
265 new natural albicidin derivatives with unambiguous identification.

266  
267

## 268 **Conclusions**

269 There are still some challenges to overcome, but both synthetic biology and metabolomics are  
270 very dynamic fields that are forging an ever-closer alliance. An example that illustrates the  
271 integral role of metabolomics in synthetic biology pipelines is the recently published multi-  
272 omics workflow to characterise strain variation in engineered *E. coli* [51]. It is certain that in  
273 coming years we will see a rapid deepening of the technical and conceptual integration of  
274 metabolic profiling methods within the Design – Build – Test cycle, and in particular the  
275 emergence of additional tools to facilitate the flow of data and insights between the analytical  
276 machinery (Test) and its users in the Design and Build stages of strain engineering.

277

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285 Advanced Pharmaceutical Ingredients (TOPCAPI).

286  
287

## 288 **References and recommended reading**

289 Papers of particular interest, published within the period of review, have been highlighted as:

290 • of special interest

291

292 •• of outstanding interest

293

294 1. Breitling, R. and E. Takano, *Synthetic biology advances for pharmaceutical production*.  
295 *Curr Opin Biotechnol*, 2015. **35**: p. 46-51.

296 2. Breitling, R. and E. Takano, *Synthetic Biology of Natural Products*. Cold Spring Harb  
297 *Perspect Biol*, 2016. **8**(10).

298 3. Ren, H., P. Hu, and H. Zhao, *A plug-and-play pathway refactoring workflow for natural*  
299 *product research in Escherichia coli and Saccharomyces cerevisiae*. *Biotechnol Bioeng*,  
300 2017. **114**(8): p. 1847-1854.

301 4. Smanski, M.J., et al., *Meeting Report for Synthetic Biology for Natural Products 2017: The*  
302 *Interface of (Meta)Genomics, Machine Learning, and Natural Product Discovery*. *ACS*  
303 *Synth Biol*, 2017. **6**(5): p. 737-743.

304 5. Toogood, H.S., et al., *Enzymatic Menthol Production: One-Pot Approach Using Engineered*  
305 *Escherichia coli*. *Acs Synthetic Biology*, 2015. **4**(10): p. 1112-1123.

- 306 6. Machas, M.S., R. McKenna, and D.R. Nielsen, *Expanding Upon Styrene Biosynthesis to*  
307 *Engineer a Novel Route to 2-Phenylethanol*. Biotechnology Journal, 2017. **12**(10).  
308 7. Zirpel, B., et al., *Engineering yeasts as platform organisms for cannabinoid biosynthesis*.  
309 Journal of Biotechnology, 2017. **259**: p. 204-212.  
310 8. Sucipto, H., et al., *Heterologous production of myxobacterial  $\alpha$ -pyrone antibiotics in*  
311 *Myxococcus xanthus*. Metabolic Engineering, 2017. **44**: p. 160-170.  
312 9. Kasuga, K., et al., *Heterologous production of kasugamycin, an aminoglycoside antibiotic*  
313 *from Streptomyces kasugaensis, in Streptomyces lividans and Rhodococcus erythropolis L-*  
314 *88 by constitutive expression of the biosynthetic gene cluster*. Applied Microbiology and  
315 Biotechnology, 2017. **101**(10): p. 4259-4268.  
316 10. Smanski, M.J., et al., *Synthetic biology to access and expand nature's chemical diversity*.  
317 Nature Reviews Microbiology, 2016. **14**(3): p. 135-149.  
318 11. Chen, P.W., M.K. Theisen, and J.C. Liao, *Metabolic systems modeling for cell factories*  
319 *improvement*. Current Opinion in Biotechnology, 2017. **46**: p. 114-119.  
320 12. Yan, C.Y., et al., *Real-Time Screening of Biocatalysts in Live Bacterial Colonies*. Journal of  
321 the American Chemical Society, 2017. **139**(4): p. 1408-1411.  
322 13. Cheng, S., et al., *Online Monitoring of Enzymatic Reactions Using Time-Resolved*  
323 *Desorption Electrospray Ionization Mass Spectrometry*. Analytical Chemistry, 2017.  
324 **89**(4): p. 2338-2344.  
325 14. Zhang, L., S. Smart, and T.R. Sandrin, *Biomarker- and similarity coefficient-based*  
326 *approaches to bacterial mixture characterization using matrix-assisted laser desorption*  
327 *ionization time-of-flight mass spectrometry (MALDI-TOF MS)*. Scientific Reports, 2015. **5**.  
328 15. AlMasoud, N., et al., *Classification of Bacillus and Brevibacillus species using rapid analysis*  
329 *of lipids by mass spectrometry*. Analytical and Bioanalytical Chemistry, 2016. **408**(27): p.  
330 7865-7878.  
331 ••16. Si, T., et al., *Profiling of Microbial Colonies for High-Throughput Engineering of Multistep*  
332 *Enzymatic Reactions via Optically Guided Matrix-Assisted Laser Desorption/Ionization*  
333 *Mass Spectrometry*. Journal of the American Chemical Society, 2017. **139**(36): p. 12466-  
334 12473.  
335 In this article, the authors show how a optically-guided MALDI-MS allows the rapid  
336 identification and recovery of desirable strains from large populations of microbial colonies.  
337  
338 17. Kompauer, M., S. Heiles, and B. Spengler, *Atmospheric pressure MALDI mass spectrometry*  
339 *imaging of tissues and cells at 1.4- $\mu$  m lateral resolution*. Nature Methods, 2017. **14**(1):  
340 p. 90-96.  
341 18. Comi, T.J., et al., *Categorizing Cells on the Basis of their Chemical Profiles: Progress in*  
342 *Single-Cell Mass Spectrometry*. Journal of the American Chemical Society, 2017. **139**(11):  
343 p. 3920-3929.  
344 19. Link, H., et al., *Real-time metabolome profiling of the metabolic switch between starvation*  
345 *and growth*. Nature Methods, 2015. **12**(11): p. 1091-1097.  
346 20. Materic, D., et al., *Monoterpene separation by coupling proton transfer reaction time-of-*  
347 *flight mass spectrometry with fastGC*. Analytical and Bioanalytical Chemistry, 2015.  
348 **407**(25): p. 7757-7763.  
349 •21. Khomenko, I., et al., *Non-invasive real time monitoring of yeast volatilome by PTR-ToF-MS*.  
350 Metabolomics, 2017. **13**(10).  
351 This study illustrates the high-throughput on-line monitoring of the volatile metabolome of  
352 microbial colonies enabled by automatic proton transfer reaction mass spectrometry.  
353  
354 22. Materic, D., et al., *Selective reagent ion-time of flight-mass spectrometry study of six*  
355 *common monoterpenes*. International Journal of Mass Spectrometry, 2017. **421**: p. 40-50.  
356 23. Jiang, G.Z., et al., *Manipulation of GES and ERG20 for geraniol overproduction in*  
357 *Saccharomyces cerevisiae*. Metabolic Engineering, 2017. **41**: p. 57-66.



- 358 24. Amiri, P., et al., *Metabolic engineering of Saccharomyces cerevisiae for linalool production*.  
359 Biotechnology Letters, 2016. **38**(3): p. 503-508.
- 360 25. Jongedijk, E., et al., *Capturing of the monoterpene olefin limonene produced in*  
361 *Saccharomyces cerevisiae*. Yeast, 2015. **32**(1): p. 159-171.
- 362 26. Covington, B.C., J.A. McLean, and B.O. Bachmann, *Comparative mass spectrometry-based*  
363 *metabolomics strategies for the investigation of microbial secondary metabolites*. Natural  
364 Product Reports, 2017. **34**(1): p. 6-24.
- 365 27. Causon, T.J. and S. Hann, *Review of sample preparation strategies for MS-based*  
366 *metabolomic studies in industrial biotechnology*. Analytica Chimica Acta, 2016. **938**: p.  
367 18-32.
- 368 ••28. Li, Z.C., et al., *Integrating MS1 and MS2 Scans in High-Resolution Parallel Reaction*  
369 *Monitoring Assays for Targeted Metabolite Quantification and Dynamic C-13-Labeling*  
370 *Metabolism Analysis*. Analytical Chemistry, 2017. **89**(1): p. 877-885.
- 371 Here, the authors present the application of parallel reaction monitoring in metabolomics, as a  
372 method to rapidly characterize the metabolic pathway reorganization observed in engineered *E.*  
373 *coli* strains.
- 374
- 375 29. Zhou, J.T., et al., *Development and Evaluation of a Parallel Reaction Monitoring Strategy*  
376 *for Large-Scale Targeted Metabolomics Quantification*. Analytical Chemistry, 2016.  
377 **88**(8): p. 4478-4486.
- 378 30. Naz, S., et al., *Development of a Liquid Chromatography High Resolution Mass*  
379 *Spectrometry Metabolomics Method with High Specificity for Metabolite Identification*  
380 *Using All Ion Fragmentation Acquisition*. Analytical Chemistry, 2017. **89**(15): p. 7933-  
381 7942.
- 382 ••31. Ohtake, T., et al., *Metabolomics-driven approach to solving a CoA imbalance for improved*  
383 *1-butanol production in Escherichia coli*. Metabolic Engineering, 2017. **41**: p. 135-143.
- 384 Using global metabolic profiling, the authors identify a non-obvious new target for enhancing  
385 the production of a desired end compound in engineered microbes.
- 386
- 387 ••32. Noguchi, S., et al., *Quantitative target analysis and kinetic profiling of acyl-CoAs reveal the*  
388 *rate-limiting step in cyanobacterial 1-butanol production*. Metabolomics, 2016. **12**(2).  
389 A combination of label-free global metabolite profiling and stable-isotope labelled fluxomics  
390 revealed new potential rate-limiting steps in a well-studied metabolic pathway.
- 391
- 392 33. Liu, L.N., X.G. Duan, and J. Wu, *Modulating the direction of carbon flow in Escherichia coli*  
393 *to improve L-tryptophan production by inactivating the global regulator FruR*. Journal of  
394 Biotechnology, 2016. **231**: p. 141-148.
- 395 ••34. Rodriguez, A., et al., *Plasmid-encoded biosynthetic genes alleviate metabolic*  
396 *disadvantages while increasing glucose conversion to shikimate in an engineered*  
397 *Escherichia coli strain*. Biotechnology and Bioengineering, 2017. **114**(6): p. 1319-1330.
- 398 A combination of metabolomics profiling and fluxomics experiments provided detailed insights  
399 into the profound metabolic consequences of the overproduction of a key precursor for an  
400 important family of natural products.
- 401
- 402 35. Lai, B., et al., *Quantitative analysis of aromatics for synthetic biology using liquid*  
403 *chromatography*. Biotechnology Journal, 2017. **12**(1): p. 1600269.
- 404 36. Teoh, S.T., et al., *A metabolomics-based strategy for identification of gene targets for*  
405 *phenotype improvement and its application to 1-butanol tolerance in Saccharomyces*  
406 *cerevisiae*. Biotechnology for Biofuels, 2015. **8**.
- 407 37. Jensen, E.D., et al., *Transcriptional reprogramming in yeast using dCas9 and combinatorial*  
408 *gRNA strategies*. Microbial Cell Factories, 2017. **16**.

- 409 38. Liu, H., et al., *High titer mevalonate fermentation and its feeding as a building block for*  
410 *isoprenoids (isoprene and sabinene) production in engineered Escherichia coli*. *Process*  
411 *Biochemistry*, 2017. **62**: p. 1-9.
- 412 39. Kirby, J., et al., *Engineering a functional 1-deoxy-D-xylulose 5-phosphate (DXP) pathway in*  
413 *Saccharomyces cerevisiae*. *Metabolic Engineering*, 2016. **38**: p. 494-503.
- 414 40. Quinn, R.A., et al., *Molecular Networking As a Drug Discovery, Drug Metabolism, and*  
415 *Precision Medicine Strategy*. *Trends in Pharmacological Sciences*, 2017. **38**(2): p. 143-  
416 154.
- 417 41. Wang, M.X., et al., *Sharing and community curation of mass spectrometry data with Global*  
418 *Natural Products Social Molecular Networking*. *Nature Biotechnology*, 2016. **34**(8): p.  
419 828-837.
- 420 42. Olivon, F., et al., *MZmine 2 Data-Preprocessing To Enhance Molecular Networking*  
421 *Reliability*. *Analytical Chemistry*, 2017. **89**(15): p. 7836-7840.
- 422 43. Olivon, F., et al., *Optimized experimental workflow for tandem mass spectrometry*  
423 *molecular networking in metabolomics*. *Analytical and Bioanalytical Chemistry*, 2017.  
424 **409**(24): p. 5767-5778.
- 425 •44. Mohimani, H., et al., *Dereplication of peptidic natural products through database search of*  
426 *mass spectra*. *Nature Chemical Biology*, 2017. **13**(1): p. 30-37.
- 427 This study illustrates the power of computational metabolomics, introducing a new algorithm  
428 for the dereplication of natural products libraries using molecular networking strategies.  
429
- 430 ••45. Nguyen, D.D., et al., *Indexing the Pseudomonas specialized metabolome enabled the*  
431 *discovery of poeamide B and the bananamides*. *Nature Microbiology*, 2017. **2**(1).  
432 Metabolite networking enabled the rapid parallel exploration of the secondary metabolome of  
433 more than two hundred ecologically diverse strains of microbes, facilitating dereplication, while  
434 at the same time identifying structural relationships between newly discovered compounds.  
435
- 436 46. Crusemann, M., et al., *Prioritizing Natural Product Diversity in a Collection of 146*  
437 *Bacterial Strains Based on Growth and Extraction Protocols*. *Journal of Natural Products*,  
438 2017. **80**(3): p. 588-597.
- 439 47. von Eckardstein, L., et al., *Total Synthesis and Biological Assessment of Novel Albicidins*  
440 *Discovered by Mass Spectrometric Networking*. *Chemistry-a European Journal*, 2017.  
441 **23**(61): p. 15316-15321.
- 442 48. Hartmann, A.C., et al., *Meta-mass shift chemical profiling of metabolomes from coral reefs*.  
443 *Proceedings of the National Academy of Sciences of the United States of America*, 2017.  
444 **114**(44): p. 11685-11690.
- 445 49. Hoffman, M., et al., *Homospermidine Lipids: A Compound Class Specifically Formed during*  
446 *Fruiting Body Formation of Myxococcus xanthus DK1622*. *Acs Chemical Biology*, 2018.  
447 **13**(1): p. 273-280.
- 448 50. Okada, B.K., et al., *Mapping the Trimethoprim-Induced Secondary Metabolome of*  
449 *Burkholderia thailandensis*. *Acs Chemical Biology*, 2016. **11**(8): p. 2124-2130.
- 450 51. Brunk, E., et al., *Characterizing Strain Variation in Engineered E.coli Using a Multi-Omics-*  
451 *Based Workflow*. *Cell Systems*, 2016. **2**(5): p. 335-346.
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455 **Table 1**  
 456 Glossary of analytical technologies

<b>Technique/Approach</b>	<b>Full Name</b>	<b>Description</b>
Metabolomics	—	The untargeted, non-biased detection and identification of all low-molecular weight compounds (metabolites) present within a biological sample or system.
MS	Mass Spectrometry	Analytical technique based on the ionisation of analytes (e.g., by DESI, MALDI, PTR or SIFT; see below), the subsequent separation of ions according to mass/charge ratio, and their detection and quantification.
MSI	Mass Spectrometry Imaging	Mass spectrometry is conducted in a spatial manner thus permitting the visualisation of the two-dimensional localisation of analytes within a sample, for example across a microbial colony growing on an agar plate.
DESI-MS	Desorption Electrospray Ionization Mass Spectrometry	Ambient ionization technique using a nebulized electrospray. Highly charged microdroplets collect analytes from the surface of the sample prior to secondary droplets carrying the analyte to the MS. This ionization technique is particularly suitable for MSI.
IM-MS	Ion Mobility Mass Spectrometry	A variant of MS, with additional separation of ions according to the time it takes for them to travel through a drift tube with a homogeneous, continuous electric field in the presence of a neutral gas. This leads to separation of ions according to size and shape (collision cross section), complementing the mass/charge information available in traditional MS
MALDI-MS	Matrix Assisted Laser Desorption Ionization Mass Spectrometry	Ionization approach whereby a matrix (an energy-absorbing small organic compound) is applied to/mixed with a sample. A laser applied to the matrix:sample mix excites the matrix molecules and leads to the generation of volatilized ions which subsequently enter the MS. This technique is suitable for MSI.
PTR-MS	Proton Transfer Reaction Mass	A soft ionization technique using an ion beam of protonated water

	Spectrometry	molecules, $\text{H}_3\text{O}^+$ , as an ion source to protonate (and thus ionize) volatile analytes. This technique permits for real-time monitoring of organic molecules in the gas phase.
SIFT-MS	Selected-Ion Flow-Tube Mass Spectrometry	Similar to PTR-MS, this soft ionisation technique uses precursor ions in the gas phase to ionize volatile analytes. The precursor ions are generated by a microwave plasma ion source, and a single ion species can be selected ( $\text{H}_3\text{O}^+$ , $\text{NO}^+$ or $\text{O}_2^-$ ) to perform as reactant ion. Neutral volatile analyte molecules react with the precursor ions and undergo ionization. This technique permits for real-time monitoring in the gas phase.
Molecular Networking	—	A computational method for MS data analysis that allows for the identification of sets of spectra from chemically related molecules ("spectral networks"), based on similarities in molecular fragmentation patterns, even in the cases when the spectra are not matched to any known compounds.

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