

ANNALS OF THE NEW YORK ACADEMY OF SCIENCES

Special Issue: Antimicrobial Therapeutics Reviews

Review

Advances in fungal chemical genomics for the discovery of new antifungal agents

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Invasive fungal infections have escalated from a rare curiosity to a major cause of human mortality around the globe. This is in part due to a scarcity in the number of antifungal drugs available to combat mycotic disease, making the discovery of novel bioactive compounds and determining their mode of action of utmost importance. The development and application of chemical genomic assays using the model yeast *Saccharomyces cerevisiae* has provided powerful methods to identify the mechanism of action of diverse molecules in a living cell. Furthermore, complementary assays are continually being developed in fungal pathogens, most notably *Candida albicans* and *Cryptococcus neoformans*, to elucidate compound mechanism of action directly in the pathogen of interest. Collectively, the suite of chemical genetic assays that have been developed in multiple fungal species enables the identification of candidate drug target genes, as well as genes involved in buffering drug target pathways, and genes involved in general cellular responses to small molecules. In this review, we examine current yeast chemical genomic assays and highlight how such resources provide powerful tools that can be utilized to bolster the antifungal pipeline.

Keywords: chemical genomics; fungal pathogens; Candida; Cryptococcus; antifungal; Saccharomyces cerevisiae

Introduction

Fungal diseases are ubiquitous in nature, threatening biodiversity and world food supplies, as well as accounting for billions of human infections each year in both developing and developed nations.¹ Over the past several decades, advances in modern medicine that rely heavily on the use of immunosuppressive drugs, including chemotherapy and transplantation surgery, as well as infections with HIV, have dramatically increased the number of individuals with compromised immunity. While most individuals suffering from mycotic infection experience relatively benign superficial symptoms, immunocompromised individuals are highly susceptible to invasive infections that ravage crucial tissues and organs.^{1,2} These infections have notoriously poor clinical outcomes, where mortality rates often exceed 50% even with therapeutic intervention.¹ Furthermore, invasive fungal infections incur the majority of the \$7.2 billion in fungal-associated direct medical costs in the United States alone, despite comprising fewer than 1% of relevant hospital visits.³ These considerable health and economic burdens are likely gross underestimates due to inadequate diagnostics and an absence of mycological surveillance.¹ In conjunction with lax antifungal stewardship that applies unnecessary selective pressure on fungal populations, it is not surprising that we are bearing witness to ever-increasing levels of drug resistance across established and emerging fungal pathogens.^{4–7}

Over 90% of fungal-related mortality is attributable to opportunistic *Candida*, *Cryptococcus*, and *Aspergillus* species.¹ *Candida albicans* is a commensal member of the human microbiota, but is also a primary causative agent of

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Figure 1. Timeline of antifungal discovery, and antifungal drug mechanisms of action. (A) The clinical introduction of antifungal agents belonging to the three major clinically used classes: polyenes, azoles, and echinocandins. (B) The mechanisms of action of antifungal drugs. Azoles exert fungistatic activity by inhibiting lanosterol 14- α -demethylase (encoded by *ERG11*), which leads to a block in ergosterol synthesis and the accumulation of toxic sterol intermediates, produced by Erg3 (left panel). Polyenes act as a fungicidal "sterol sponge" by forming extra-membranous aggregates that extract ergosterol from lipid bilayers (middle panel). Fungal cell walls are composed of (1,3)- β -D-glucan covalently linked to (1,6)- β -D-glucan, as well as chitin and mannan. Echinocandins prevent the synthesis of (1,3)- β -D-glucan by inhibiting the (1,3)- β -D-glucan synthase (encoded by *FKS1* in *C. albicans* and by both *FKS1* and *FKS2* in *S. cerevisiae*); this results in a loss of cell wall integrity (right panel). Adapted, with permission, from Ref 134.

life-threatening bloodstream infections.8,9 In addition, the unprecedented, global rise of non-albicans Candida species with problematic levels of intrinsic and acquired drug resistance is deeply concerning. Since its sudden global emergence in 2009, Candida auris has been implicated in numerous nosocomial outbreaks, with 93% of C. auris clinical isolates exhibiting resistance to the most widely deployed antifungal class, the azoles, and 4% of strains being recalcitrant to all available antifungal classes.^{6,10} Similarly, high levels of antifungal resistance have been documented for Candida glabrata, which currently represents the second most commonly isolated Candida species in the United States and Europe.^{11,12} Cryptococcosis often manifests as devastating central nervous system infections in immunocompromised patients and has associated mortality rates approaching 70% in endemic regions.¹³ Cryptococcus neoformans and Cryptococcus gattii are the predominant culprits, with C. gatti also capable of causing disease in immunocompetent hosts.^{14,15} Lastly, Aspergillus fumigatus has a universal environmental presence and is responsible for over 200,000 reported invasive infections annually with burgeoning azole resistance reported.^{1,16}

Currently, there are only three major classes of clinically used antifungal drugs to treat systemic infection, which interfere with just two central fungal processes (Fig. 1).^{11,17} Azoles are the most widely deployed antifungals owing to their broad-spectrum activity, favorable safety profile, and oral bioavailability.^{18,19} They target lanosterol $14-\alpha$ -demethylase, diverting ergosterol biosynthesis toward the production of a toxic sterol intermediate that exerts a severe cell membrane stress, blocking further cell growth and division (Fig. 1B).^{18,19} Unfortunately, their fungistatic action and widespread, prophylactic use in both medicine and agriculture poses strong directional selection pressure for the evolution of resistance.4,7,16 Polyenes, such as amphotericin B, directly bind and remove ergosterol from the cell membrane acting as a sterol sponge (Fig. 1B).²⁰ Although acquired resistance in clinically relevant contexts remains rare,^{4,7,21} polyenes are instead associated with undesirable host nephrotoxicity. Thus, amphotericin B is usually deployed as a last resort therapy against cryptococcal meningitis²² and multidrug-resistant *Candida* infections.²³ Finally, the echinocandins are the only novel drug class approved for clinical application in the past two decades.¹¹ They noncompetitively inhibit 1,3- β -D-glucan synthase, depriving the fungal cell wall of an integral polymeric component, leading to the induction of cell wall stress and eventually cellular death (Fig. 1B).^{4,7}

The stagnant discovery of novel antifungal compounds over the past several decades can be attributed to a variety of factors. First and foremost, the number of fungal-specific cellular pathways that can be therapeutically targeted is restricted due to the eukaryotic similarities between fungi and humans, compared with more evolutionarily distant microbes, such as bacteria.18,19 In addition, an estimated 80% of molecules with published fungicidal activity are not pursued due to inherent unsuitable properties, namely promiscuous functionality, as well as a lack of whole-cell bioactivity due to difficulties with permeating across the fungal cell wall and membrane.^{18,24} Poor target specificity and lack of whole-cell bioactivity also contribute to high attrition rates during preclinical development. Finally, the scientific complexities of antifungal development are exacerbated by a chronic dearth of research initiatives and resource investment. Yet, despite these challenges, scientific advancements have offered great hope for the discovery and development of novel antifungal agents. Phenotype-based identification of bioactive molecules, which prioritizes potent activity within the context of intact fungal cells, can now be coupled with a vast array of functional genomics resources to enable subsequent target deconvolution.^{25–28} In this review, we explore current chemical-genomic approaches developed for Saccharomyces cerevisiae, as well as those resources available in clinically important fungal pathogens, emphasizing their potential to revitalize antifungal drug discovery.

The power of yeast genetics

Since its genome sequence was assembled almost 25 years ago,²⁹ *S. cerevisiae* has prevailed as the *de facto* eukaryotic model system, guiding our understanding of the eukaryotic cell and serving as an

invaluable resource of genomic tools and reagents. Systematic genetic and phenotypic analyses in this species were enabled by a consortium of scientific laboratories that developed a comprehensive collection of deletion mutants, where each of the \sim 6000 yeast open reading frames was replaced with a *kanMX* dominant, drug-resistance marker.³⁰ This strategy also encompassed the inclusion of strain-specific molecular sequences: two unique 20 base-pair oligonucleotides that are flanked by universal priming sequences. These strain-specific barcodes surround each kanMX drug-resistance marker and serve as a unique molecular identifier for each strain in the S. cerevisiae genome.³¹ This systematic endeavor defined the set of ~ 1000 S. cerevisiae genes essential for viability in laboratory growth conditions and generated a set of ~5000 viable haploid deletion mutants.^{30,32} Notably, the essential gene set is not only contingent on standard growth conditions but also strain background, which has implications for drug development and how researchers might select potential drug targets. Importantly, this strategy motivated several other laboratories to develop additional genomewide tools to enable proteomic studies, such as the yeast comprehensive temperature-sensitive mutant collection,^{33,34} the yeast titratable-promoter collection of essential genes,^{35,36} the tandem affinity purification (TAP-tagged) collection,³⁷ the GFP (green fluorescent protein) collection,³⁸ and genome-scale two-hybrid resources.³⁹ More recently, complementary strain collections were constructed in which subsets of the \sim 1000 essential yeast genes were individually altered to produce conditional or hypomorphic alleles with the potential to be assayed at a semipermissive state, enabling systematic analysis of essential genes.⁴⁰ Collectively, these mutant collections have enabled the development and application of high-throughput methodologies that have revealed incredible complexities inherent of eukaryotic biological systems.

The functional redundancy and extensive buffering within the eukaryotic cell became evident once it was revealed that only $\sim 17\%$ of genes in the *S. cerevisiae* genome are essential under standard laboratory conditions.³⁰ For decades prior, geneticists had investigated relationships between gene pairs, mapping out genetic interactions on a case-by-case basis. A genetic interaction between two genes is observed when a phenotype caused

by a mutation in one gene depends on a mutation in another gene such that the combined effect deviates from expectation based on the individual effects, with synthetic lethality representing the extreme case of a negative genetic interaction (Fig. 2A and B).41-43 For example, one of the first examples of synthetic lethality was described with loss-of-function mutant alleles of both TUB1 and TUB3, both of which encode α -tubulin.⁴⁴ With the completion of the yeast deletion collection, researchers began systematically mapping all genetic interactions in S. cerevisiae. To do so, synthetic genetic array (SGA) was developed as an automated method that combined arrays of gene deletion mutants with robotic manipulation for high-throughput construction of combinations of mutant alleles and identification of genetic interactions.⁴⁵ In its first two applications, SGA methodology crossed at first eight followed by \sim 130 gene-specific query mutant strains to the \sim 5000 haploid deletion mutants resulting in complex networks of thousands of synthetic lethal or synthetic sick interactions.45,46 Recently, the combination of SGA with a genome-scale colony size-scoring methodology enabled the assessment of growth defects associated with approximately 18 million yeast gene deletion pairs. These large-scale studies measured the fitness of single and double mutants to identify nearly 550,000 negative genetic interactions and 350,000 positive genetic interactions, enabling the assembly of the first complete genetic interaction network for any organism.^{40,47}

The mapping of genetic interactions in S. cerevisiae highlighted several critical observations with important implications for the development of novel antifungal therapies.^{45,47} While only ~ 1000 genes in S. cerevisiae are essential for viability,³⁰ the identification of 550,000 negative genetic interactions between gene deletion pairs^{40,47} suggests the potential for combatting fungal infections may come from targeting multiple cellular nodes that together result in a lethal or sick phenotype (Fig. 2B).^{18,48} For example, in S. cerevisiae, FKS1 and its paralog FKS2 encode the biosynthetic enzyme for (1,3)- β -D-glucan synthesis and the molecular target of the echinocandins. While FKS1 and FKS2 are synthetic lethal, in keeping with echinocandin efficacy, FKS1 is also synthetic lethal with CHS3, a chitin synthase required for the synthesis of the cell wall component chitin,47 and inhibitors of chitin synthases, such as nikkomycin, are synergistic with caspofungin against numerous fungal pathogens.^{49,50} While this example highlights genetic interactions as determinants of the efficacy of compound combinations, the complexities of chemical action and genetic network density may preclude the prediction of synergism on a genome-wide scale.⁵¹ Thus, additional factors must be employed in order to identify effective antifungal combinations.

Chemical genomics in S. cerevisiae

The creation of the yeast deletion collection with strain-specific molecular barcodes enables the quantification of individual strains in a mixed population.^{31,52} Consequently, this approach has been widely used over the past two decades to investigate interactions between genes and small molecules.^{32,53-56} Such chemical-genetic assays are based on the concept that modifying the expression of a compound target (or the expression of other factors involved in a process that is targeted by the compound) alters the amount of compound required to effectively inhibit that target (see Figs. 3) and 4 below). This principle has been exploited at genome scale, allowing unbiased screening to determine a molecule's mode-of-action in whole yeast cells. Benefits of the pooled chemogenomic format include the consumption of relatively little compound, which may be a limiting factor in terms of quantity and expense; the efficiency of processing pooled samples compared with profiling thousands of individual mutants; as well as the minimization of technical variation.26,27,52,57,58 To further enhance the utility of this approach, a set of donor strains, called Barcoders, was constructed that allows unique barcode sequences to be transferred to any S. cerevisiae strain collection in a rapid and cost-effective manner, enabling the application of parallel pooled approaches to a wide variety of complex bioassays.59

Haploinsufficiency, homozygous, and haploid deletion profiling

Haploinsufficiency profiling (HIP) operates under the principle that deletion of one copy of a target gene in a diploid organism confers hypersensitivity to chemical inhibition.⁶⁰ Drug-induced loss of activity of the remaining gene product emulates a complete gene deletion and is exhibited as a



Figure 2. Application of genetic-genetic and chemical-genetic interactions for compound to target annotation. (A) A genetic interaction between two genes is observed when a phenotype caused by a mutation in one gene is exacerbated by a mutation in another gene. A negative genetic interaction occurs if the observed fitness of the double mutant is less than the double mutant fitness expected from a multiplicative model. A positive genetic interaction occurs if the observed fitness of the double mutant is greater than the double mutant fitness expected from a multiplicative model. (B) Mutations in genes residing in parallel pathways typically culminate in a negative genetic-genetic interaction (green box). Genetic inhibition of one pathway combined with pharmacological inhibition of a parallel pathway typically results in a negative chemical-genetic interaction (yellow box). (C) Comparing genetic-genetic and chemical-genetic interaction profiles can facilitate functional interpretation of chemical-genetic screening data for a given compound.

quantifiable growth defect.⁵⁵ This is particularly true when the compound target is an essential (or highly critical) gene. In this case, the heterozygous strain likely shows no growth defect in untreated conditions but becomes inviable and drops out of a population in the presence of the compound that has effectively inhibited the function of the remaining target protein. HIP enables all deletion strains in the S. cerevisiase genome to be pooled together in order to determine competitive growth differences in a single genome-wide experiment by quantification of the barcodes traditionally through microarray and currently via high-throughput sequencing (Fig. 3).^{26,55,60} This experimental approach provides a relative rank order of hypersensitive mutants, as well as information as to the genes whose perturbation is associated with significant phenotypic effects.⁵⁵ Initial applications of HIP proved the accuracy of the experimental approach, as mutants of genes encoding well-characterized targets of antimicrobial agents were readily identified as the most sensitive from the pool of ~6000 heterozygous deletion strains.^{55,60,61} In addition, heterozygous deletion mutants in gene encoding products that belonged to related signaling or metabolic pathways, or factors that reduced compound availability, were often shown to be hypersensitive.^{26,55} For example, S. cerevisiae pooled screening with methotrexate identified hypersensitive mutants, including for the reported drug target dihydrofolate reductase gene, DFR1, as well as FOL1 and FOL2, which encode upstream components in the folic acid biosynthetic pathway.55 The utility of HIP in uncovering novel mechanisms of action for countless small molecules has been repeatedly demonstrated, including two independent studies that implicated ribosomal RNA processing as the mechanism of 5-fluorouracil,^{55,61} an antitumoural agent widely considered to act via inhibition of the DNA synthesis enzyme thymidylate synthetase.⁶² Furthermore, HIP identified new chemical probes targeting septin, actin, and tubulin,⁵⁶ as well as numerous alternative eukaryotic targets for diverse psychoactive drugs.63

While these examples highlight the potential of HIP in uncovering the mechanism of small molecules, not all genes display haploinsufficiency and increased sensitivity upon a 50% reduction of dosage. Furthermore, gene expression in select heterozygous deletion mutants may in fact be greater than the expected 50% due to transcriptional upregulation, which can compensate for reduced copy number.64 The decreased abundance by mRNA perturbation (DAmP) strategy is an alternative to HIP that uses antibiotic resistance cassettes to systematically disrupt the 3'-untranslated region of essential genes in haploid yeast, destabilizing the corresponding transcripts to an estimated 10% of wild-type expression.^{65,66} This methodology was used to generate a barcoded collection encompassing ~1400 strains, including approximately 90% of all essential genes.^{59,65,66} This analogous approach is preferable to simply elevating drug concentrations in heterozygous deletion mutants, which may cause loss of target specificity and thus generalized cellular toxicity.26,55

Homozygous deletion profiling (HOP) or haploid deletion profiling (HAP) is conceptually and experimentally similar to HIP but employs full deletion of nonessential genes in either diploid (HOP) or haploid (HAP) S. cerevisiae strains (Fig. 3). These approaches provide complementary and powerful assays in order to probe the mechanism of action of compounds that lack an essential protein target, and instead target cellular factors, such as DNA or lipids, or for those compounds that have redundant protein targets.^{26,67} For example, HIP of the DNAintercalating agent cisplatin proved unsuccessful at identifying a putative target as there was no single protein in the cell for which genetic reduction would confer sensitivity to a compound that binds DNA. However, parallel HOP analysis uncovered numerous DNA repair factors, supporting DNA itself as the target.⁵⁵ Although these approaches can be performed in either haploid or diploid gene deletion libraries, diploid strains present the advantage of being less impacted by secondary-site mutations, which have the potential to confound data interpretation.⁵³ HOP/HAP also allows for the identification of genes in target-related pathways or detoxification processes, including metabolism and efflux, which buffer the cellular response to chemical stress.^{26,64,68} Whereas HIP analysis often produces fewer than 10 genes of interest, HOP/HAP assays commonly identify in the tens-to-hundreds of significantly hypersensitive mutants.

The complementary application of these chemical-genetic approaches has proved invaluable to deciphering the mode of action of bioactive agents in the context of a living eukaryotic cell. One of the first large-scale chemogenomic screening efforts performed 1144 full-genome HIP experiments and 418 HOP experiments.⁵⁴ This study revealed a phenotype for 97% of genes in S. cerevisiae, suggesting that virtually the entire yeast genome is conditionally essential and therefore accessible to inhibition with small molecules;⁵⁴ this is a critical observation considering the difficulty in identifying novel single-agent antifungals. A more recent large-scale analysis profiled 3250 small molecules and identified 317 compounds that specifically targeted the function of 121 genes. This work suggested that the cellular response to small molecules is limited by a network of 45 major chemogenomic signatures, and coupled such cellular processes to specific chemical moieties.⁵⁶ Finally, recent advances in chemical-genetic screening platforms coupled with comprehensive sequence databases have further expanded our ability to profile compounds in a rapid and systematic manner, such that 13,524 molecules were investigated in a single study.⁶⁹ This was achieved by using a diagnostic set of viable yeast gene deletion mutants spanning all major biological processes in a drug-sensitized genetic background, coupled with a highly multiplexed (768-plex) barcode sequencing protocol. For data analysis, the authors also generated a computational platform to functionally annotate compounds to specific biological processes and pathways.⁶⁹⁻⁷¹

With recent technological advances that have enabled large-scale data generation, sophisticated computational platforms have proven imperative for the analysis of complex chemical-genetic interactions. Correlations between distinct chemicalgenetic fitness profiles can be used for "guilt-byassociation" to infer the mode of action of novel compounds, as compounds exhibiting similar chemical-genetic interactions tend to target similar processes (Fig. 2C).^{56,67,69,72} Likewise, the phenotypes of uncharacterized targets with known small molecules can glean insight into their biological roles. To make these comparisons, chemicalgenetic profiles can be mapped onto the global genetic interaction profile similarity network.⁴⁷ Many of these genetic, chemical, and chemicalgenetic relationships can be explored using online resources, such as the publicly accessible HIP-HOP chemogenomics database⁵⁶ and the MOSAIC database.69,71

Multicopy suppression profiling

Multicopy suppression profiling (MSP) is an orthogonal gene dosage-based approach built upon the concept that target gene overexpression enables increased tolerance to drug exposure (Fig. 4).⁵³ Traditional suppressor screens based on such ideology have been employed for decades. Such studies typically use a high-copy plasmid library carrying randomly generated yeast genomic inserts to identify genes that, when overexpressed, confer resistance. Notably, these screens involve cumbersome plating techniques and clone characterization where plasmids are isolated from resistant colonies and sequenced in order to identify the gene(s) responsible for conferring the resistance phenotype. This approach was employed to identify the target of tunicamycin as Alg7.73 With the creation of advanced functional genomic resources, this approach was adapted in order to culture pools of strains in liquid medium in a manner analogous to the HIP assay.⁵³ One of the first examples of MSP being employed in a pooled manner involved a high-copy plasmid collection containing yeast genomic DNA fragments with genes expressed from native promoters (Table 1). This library was screened in whole cells at high concentrations of compounds that inhibited a wildtype control by approximately 90%. Under such conditions, only one or a few resistant strains were selected from the population.⁶⁸ Plasmids were then isolated from resistant cells, inserts were amplified by PCR and hybridized to a microarray carrying probes complementary to each yeast open reading frame in order to determine gene abundance in compound-treated relative to an untreated control pool.⁵³ This approach correctly identified Dfr1, Erg11, and Tor1 as the targets of methotrexate, fluconazole, and rapamycin, respectively.⁵³ Similarly, the Yeast Genome Tiling collection contains overlapping fragments of the yeast genome, which are each approximately 10 kb in size, cloned into high-copy vectors (Table 1).74 The ends of each insert of this library have been sequenced, and the plasmids organized in a tiling fashion across the yeast genome, ensuring near-saturation (97.2%) coverage of the yeast genome. In both examples, each insert contains multiple genes, and thus once a fragment that confers resistance is identified, the specific gene target must be cloned and its effect on the resistance confirmed.





Figure 3. Reduced gene dosage chemical-genomic assays. Haploinsufficiency profiling (HIP), homozygous deletion profiling (HOP), and haploid deletion profiling (HAP) operate under a common principle. HIP employs heterozygous deletion mutants of essential or nonessential genes in a diploid for reduced gene dosage. HOP utilizes homozygous deletion mutants in a diploid parent and HAP utilizes haploid deletion mutants generated in a haploid parent to abolish the expression of nonessential genes. Individual strains within each genome-wide library are tagged with two unique DNA barcodes, upstream and downstream (BC1 and BC2), that permit simultaneous analysis within a single pool. Deletion libraries are grown competitively in the absence and presence of a compound of interest. Genomic DNA is isolated after a duration of pooled growth, and PCR amplification of strain-identifying barcodes is performed using universal primers for the upstream or downstream barcodes. High-throughput barcode sequencing and normalization to the untreated pool is used to quantify strain representation.

	MoBY-ORF library	Yeast tiling collection	Galactose inducible libraries	Random genomic fragment libraries
PCR free		\checkmark		
Promoter type	Native	Native	Inducible	Native
Plasmid copy number	Low	High	High	High
Barcoded				
Percent genome coverage	90%	97.2%	80%	$\sim 85\%$
Average fragment size	2 kb	10 kb	1.5 kb	5 kb
Number of genes per fragment	1	4-6	1	2-3
References	69	66	67, 68	48

Table 1.	Features of	plasmid libraries	used for multicopy	suppression profiling

Several alternative libraries offer advantages over the traditional randomly generated yeast genomic inserts collections described above. Two such resources consist of thousands of yeast strains, each carrying a vector harboring a single yeast open reading frame under the expression of the GAL1 promoter (Table 1).^{75,76} A proof-of-principle experiment with one of these inducible collections successfully identified vectors carrying TOR1 as the target of rapamycin.⁷⁵ Finally, the generation of the molecular barcoded yeast (MoBY) ORF collection, which comprises over 90% of the yeast genome, was completed in order to further enhance the utility and efficiency of this experimental approach (Table 1).77 The MoBY ORF library is constructed with a centromere-based vector where each gene is tagged with two unique DNA barcodes, similar to the yeast deletion collection.⁷⁷ Once pools of strains are cultured in the presence of compound, resistant strains are identified by amplifying the strain-specific barcodes and employing microarrays or high-throughput sequencing for strain identification (Fig. 4). This approach has been successful at predicting the mode-of-action of multiple molecules.⁷⁷ For example, chitosan oligosaccharide was known to exert fungal cell membrane stress, similar to the azoles and polyenes.⁷⁸ However, the MoBY-ORF collection determined that overexpression of the Ras superfamily GTPase ARL1 confers resistance to chitosan oligosaccharide, highlighting a previously unappreciated cellular target.79

An additional application of the MoBY-ORF library is the identification of recessive genes responsible for drug resistance (Fig. 4).⁷⁷ After a recessive drug-resistant mutant is identified, it can be transformed with the MoBY-ORF collection and complementation by one or more wild-type alleles

from the collection that restores drug sensitivity, enables the identification of the gene harboring the recessive resistance mutation.⁷⁷ This assay enables the identification of drug targets in cases where the compound must interact with another protein to become toxic. Complementation of recessive drug-resistant alleles can also be used to systematically uncover general and specific resistance mechanisms. For example, this approach was employed to further define the mechanism of action of the natural product theopalauamide.⁷⁷ Theopalauamide-resistant mutants were transformed with the MoBY-ORF collection, and the clone harboring MVD1, a gene encoding an essential enzyme in an early step of the ergosterol biosynthesis pathway, was observed to restore drug sensitivity.⁷⁷ The theopalauamide-resistant strain was confirmed to contain a single amino acid substitution within the active site of Mvd1 and subsequent analyses indicated that theopalauamide binds to ergosterol, defining a novel class of sterolbinding compounds.77

As is the case with all experimental approaches, MSP has certain limitations. For example, overexpression of a target gene may not impart an observable drug resistance phenotype if the encoded protein product resides within a larger complex with distinct stoichiometric requirements.^{26,68} Increased dosage of certain genes may also itself interfere with cellular fitness, as an estimated 15% of yeast proteincoding genes are deleterious upon overexpression.⁷⁶ The use of centromere-based vectors for the MoBY-ORF collection minimizes overexpression-induced toxicity; however, it is certainly a limitation for other overexpression collections. In addition, an exceptionally sensitive drug target expressed at wild-type levels could mask the suppressor effects of another



Figure 4. Multicopy suppression profiling. The MoBY ORF library is constructed such that each gene is tagged with two unique DNA barcodes, similar to the yeast deletion collection. This collection can be transformed into a wild-type strain (left panel) or a haploid drug-resistant strain that harbors a recessive mutation (right panel). The resulting pools can be grown with a compound of choice. Plasmid DNA is isolated. Strain-specific barcodes are amplified using plasmid primers that flank each insert. High-throughput barcode sequencing and normalization to the untreated pool is used to determine strain representation. In a wild-type background, increased expression of a compound's target gene typically confers enhanced fitness in the presence of the compound relative to other genes. In a drug-resistant background that contains a recessive resistance mutation, a plasmid with significantly reduced barcode counts identifies the gene that harbors the mutation responsible for resistance to the compound of interest.

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less sensitive overexpressed target, which would therefore escape detection.⁶⁰ Finally, drug pumps or other indirect targets may dominate in the set of strains resistant to compound. Creating similar overexpression libraries in diverse drug pumpdeficient mutants can alleviate this challenge.⁶⁸

Variomics

Several years ago, an alternative to gene dosagebased approaches was reported for the systematic discovery of drug target genes, as well as resistance genes, in S. cerevisiae.⁸⁰ This experimental approach involved a genome-wide compendium of mutant libraries constructed by high complexity random mutagenesis. The allelic variants are carried on low-copy centromeric plasmids, with 5847 available sets representing 90% of the S. cerevisiae genome.⁸⁰ Each variomic library contains $>2.0 \times 10^5$ independent alleles, including both single and multiple mutated variants, to ensure high genetic diversity to maximize the likelihood of mutations spanning all encoded amino acid residues.⁸⁰ Similar to deletion and overexpression profiling, pooled yeast growth and variant allele quantification through high-throughput barcode sequencing are performed with follow-up validation of individual resistant mutants. This variomics tool was first tested against rapamycin, validating its known target genes FPR1, TOR1, and TOR2, as well as the previously described resistance gene NPR1.80 Notably, the TOR1 and TOR2 resistance alleles all incorporated mutations within their known drug-binding domain.⁸⁰ Both heterozygous and haploid variomic libraries have been constructed to determine if different types of mutations are able to confer resistance when the genetic background contains a wild-type copy of the desired allele. It was observed that haploid variomic pools tend to contain less genetic diversity at later time points during selection, given that any resistant mutations must also retain viability.⁸¹ By contrast, variomic libraries expressed in heterozygous deletion mutants can in principle achieve separation-of-function. This library enabled the identification of dfr1 hypomorphic alleles in the diploid state, which modulate methotrexate resistance, as well as previously characterized dominant dfr1 mutations in the haploid background.⁸¹

A strength of the variomics approach is that it defines key amino acid residues modulating compound interactions and selectivity, providing further resolution into drug-target engagement. As well, mutant allele preconstruction obviates the reliance on spontaneous mutation rates, which can otherwise be inconsistent. However, spontaneous resistant mutants may still emerge during selection, leading to false positives and demanding functional validation.⁸⁰ Finally, variomics-based analysis is unable to delineate resistance mechanisms that involve multiple genes.⁸⁰

Ultimately, the complementary approaches highlighted above are best applied in parallel to present a more powerful strategy to elucidate compound mode of action. Congruent results lend higher confidence to target identification. For instance, mutants of the targets of fluconazole, rapamycin, and methotrexate, although consistently hypersensitive in HIP assays with the cognate inhibitor, are also not always the most hypersensitive strains, but the corresponding target genes also confer resistance when overexpressed in MSP analysis.⁵³

Chemical genomics in fungal pathogens

While the direct extension of findings from cerevisiae chemical-genetic analyses S. to pathogenic species, like C. albicans, has been fairly successful in determining the mode of action of antifungal molecules, significant limitations regarding genetic comparisons have also arisen due to phylogenetic divergence.⁸² In particular, fungal pathogens harbor mechanisms and pathways required for virulence that either do not exist in the model yeast or have been substantially repurposed over evolutionary time.83 Moreover, the list of probable drug targets in fungal pathogens is predicted to be even greater than the druggable *S. cerevisiae* genome.^{84,85} Thus, the ability to dissect virulence attributes and drug responses in these organisms in a comprehensive manner requires the development of genetic resources directly in the fungal pathogens. Fortunately, recent advances in the generation of functional genomic tools in pathogenic fungi have enabled experimentation in the pathogens of interest.^{82,83,86,87} Despite the challenges of genetic intractability and cryptic life cycles, the availability of complete or draft genome sequences in pathogenic fungal species,88-90 coupled with advances in genetic manipulation,91-93 has enabled progress with the production of comprehensive mutant collections and permitted the systematic discovery of compound-target connections.

Chemical genomics in Candida

Thus far, the most comprehensive chemical-genetic resources among pathogenic fungi have been generated for C. albicans. Two primary functional genomics tools have been instrumental in advancing our understanding of gene function in this species. First, the C. albicans double barcoded (DBC) heterozygous deletion library covers ~90% of the C. albicans genome.⁸² Similar to the S. cerevisiae heterozygous deletion collection, the DBC library also includes strain-specific molecular sequences that are flanked by universal priming sequences. These strain-specific barcodes surround a HIS3 auxotrophic marker and serve as unique molecular identifiers for each strain in the C. albicans genome.⁸² Second, building upon the DBC collection, a library of C. albicans mutants was generated to facilitate large-scale functional analysis of its genome. To do so, a gene replacement and conditional expression (GRACE) strategy was employed where a heterozygous deletion mutant was transformed such that the expression of the remaining allele was conditionally controlled by replacing the native promoter with a tetracycline-repressible promoter.86 This GRACE collection currently consists of ~2400 mutants representing approximately \sim 40% of the *C. albicans* genome. Furthermore, this methodology has enabled a direct comparison of phenotypes under nonrepressing and repressing conditions in order to define those genes essential for C. albicans growth under laboratory conditions, as well as *in vivo*.^{86,94} Defining the compendium of essential genes in C. albicans is imperative to the elucidation of novel antifungal targets.

Additional genomic resources have expanded the repertoire of experimental approaches to interrogate gene function in *C. albicans*. The transposon insertion-based TagModule collection is another open-access, barcoded set of heterozygous deletion mutants, which comprises 59% of predicted *C. albicans* open reading frames.⁹⁵ This collection was used to identify novel regulators of filamentous growth and new genes involved in essential processes, as well as verify the target of brefeldin A as Sec7.⁹⁵ This TagModule toolkit is universal, meaning that it can be applied to diverse microorganisms to support antimicrobial research. As well, a barcoded homozygous deletion library spanning 11% of annotated *C. albicans* coding genes was generated in order to profile genes important for growth, morphogenesis, virulence, and commensalism.^{96,97} Several resources have also been developed for focused gene overexpression studies in *C. albicans*,^{98,99} albeit with limited application outside of functional annotation of specific genetic networks.

In comparison, there is a dearth of existing genetic tools in non-*albicans Candida* species. The haploid genomes of *C. glabrata* and *C. auris* are refractory to HIP, though a barcoded deletion mutant library covering 12% of the *C. glabrata* genome was recently employed to identify genes that influence azole and echinocandin susceptibility.¹⁰⁰ Likewise, recent completion of annotated *C. auris* genome assemblies¹⁰¹ and advances in CRISPR-Cas9 technology, particularly the use of RNA-Cas9 complexes for expression-free gene editing systems in *C. glabrata* and *C. auris*,⁹¹ can be envisioned as guiding large-scale loss-of-function mutant collections and drug target determination in the near future.

Chemical genomics in Cryptococcus

Genetic tool development for C. neoformans can be considered arduous, in large part due to inefficient transformation that is achieved by either electroporation or the bombardment of DNA-coated particles (biolistic transformation), low homologous recombination rates, as well as complex genomic structural features.^{90,102,103} However, the development of cloning-free fusion PCR methods for the addition of long homology arms for gene targeting, optimization of transformation parameters, the development of multiple dominant drug-selectable marker systems, and the optimization of rapid screening for genotyping transformants made it feasible to construct large numbers of gene deletions in C. neoformans.¹⁰⁴ Specifically, an open-access library of 1201 signature-tagged deletion mutants, biased for genes lacking S. cerevisiae orthologs, was constructed.¹⁰⁵ Each deletion mutant harbored one of 48 DNA oligonucletoide barcodes that enabled quantitative detection of each mutant in a pooled sample.^{104,105} Furthermore, this collection along with another barcoded gene deletion library¹⁰⁵ that together encompassed 1448 gene knockouts, was used to generate a chemogenomic atlas where each homozygous deletion mutant was cultured in the presence of one of 439 small molecules.⁸⁷ Eightythree percent of the mutants were found to be associated with chemical phenotypes (either hypersensitive or resistant), and each compound induced one or more genetic responses. Importantly, a comparison of these results with large-scale studies in S. cerevisiae^{54,67} highlighted that chemical-genetic profiles from C. neoformans were largely distinct from those in S. cerevisiae, emphasizing the relevance of pathogen-focused studies.⁸⁷ In particular, chemical-genetic screens with this deletion mutant collection were pivotal in determining that the azole synergizing agent dicyclomine increases cell permeability and decreases nutrient uptake in C. neoformans.¹⁰⁶ Additional gene deletion collections have also been generated for functional genomic profiling, including a 322 signature-tagged gene deletion collection encompassing 155 putative transcription factor genes, as well as 264 signaturetagged gene deletion mutants for 129 putative kinases.^{107,108} These collections were profiled to unveil key regulators of virulence and drug resistance in order to map potential anticryptococcal targets to be exploited for therapeutic development. Finally, the genome editing strategy transient CRISPR-Cas9 coupled with electroporation system (TRACE) was recently developed that dramatically improves the efficiency of targeted mutagenesis in the Cryptococcus genome,^{109,110} thus providing great promise for the generation of future mutant collections in this species.

Chemical genomics in Aspergillus

The major obstacles to targeted genetic manipulation in A. fumigatus are poor homologous recombination and prolific genetic exchange via vegetative cell fusion.¹¹¹ To date, systematic mutant collections for chemical-genetic fitness profiling are lacking; however, the outlook is bright with improvements in gene deletion and overexpression methods. Transposon insertion mutagenesis, as well as parasexual genetics, has been deployed to identify essential A. fumigatus genes.^{112,113} Moreover, a conditional promoter replacement strategy was implemented to distinguish and prioritize essential and putative antifungal drug target genes among 54 genes with essential C. albicans or S. cerevisiae orthologs.¹¹⁴ This included the generation of a conditional ALG7 mutant, which was confirmed to elicit specific hypersensitivity to its known inhibitor tunicamycin, upon genetic depletion in a whole cell.¹¹⁴ More recently, a study demonstrating minimal induced cell fusion under controlled culture conditions served as proof-of-principle for barcode-free, high-throughput, competitive fitness profiling in A. fumigatus.¹¹⁵ A mini-library of 46 heterozygous deletion strains of known drug target genes and/or essential genes was created by a rapid allelic replacement technique, with cyp51A and arf2 mutants rediscovered as the only significantly hypersensitive outliers for their cognate inhibitors itraconazole and brefeldin A, respectively.¹¹⁵ In addition, multiple groups have recently reported advances with CRISPR-Cas9 to promote homologous recombination for targeted gene disruption and loss-offunction studies in Aspergillus species.^{92,116,117}

Characterization of novel antifungals using chemical genomics

With the vast array of functional genomics resources available in S. cerevisiae, C. albicans, and C. neoformans, it is not surprising that the application of chemical-genomic approaches for antifungal mode of action analysis is expanding. For example, the natural product parnafungin was shown to display potent and broad spectrum activity against diverse fungal pathogens by inhibiting fungal poly(A) polymerase,¹¹⁸ and the antifungal plant defensin, RsAFP2, was demonstrated through HIP to target the C. albicans glucosylceramides, key components of the fungal cell wall.¹¹⁹ Similar research identified structurally related synthetic molecules that induced hypersensitivity in an OLE1 heterozygous mutant, implicating the biosynthesis of unsaturated fatty acids as a potential antifungal target.¹²⁰ In addition, synthetic compounds with antifungal activity against C. neoformans were shown to target sphingolipid biosynthesis using S. cerevisiae HIP-HOP to elucidate mode of action.¹²¹ Finally, recent studies unveiled great potential of targeting either glycosylphosphatidylinositol precursor biosynthesis,¹²² or the fungal casein kinase Yck2,¹²³ through their use of C. albicans HIP to identify the mechanism of action of novel bioactive compounds.

In addition to the examples highlighted above, several studies have also leveraged high-throughput approaches to determine the cellular targets of diverse compounds that display antifungal activity. Two such studies performed high-throughput screens to identify antifungal potentiators and then characterized the mode of action of top candidates using *S. cerevisiase* chemical genomic resources, implicating effects on membrane permeability or sphingolipid metabolism.^{124,125} Furthermore, efforts from industry to screen ~1800 natural product extracts identified and characterized a number of natural products with antifungal activity, including: yefafungin, which targets the fungal-specific translation initiation factor *YEF3*; campafungin, predicted to inhibit adenylate cyclase activity and/or cAMP regulation; and fellutamides C and D, which were predicted to inhibit the fungal proteasome.¹²⁶

One final alternative application of fungal chemical genomic resources to combat fungal infection is to identify genes and genetic networks important for fungal virulence and identify virulence factors targeted by novel small molecules. Targeting fungal virulence provides a complementary approach to the development of antifungal agents, as the goal is to occlude the ability of a microbe to cause harm to its host. Targeting virulence factors offers many benefits, including expanding the repertoire of antifungal targets, minimizing effects on the host mycobiome, and reducing selection pressure for the evolution of drug resistance.^{127,128} One such virulence trait in C. albicans is the ability to transition between yeast and filamentous morphologies. This developmental transition is not only imperative for the establishment of systemic infection,^{96,129} but is also required for the formation of drug-resistant biofilms on surfaces, such as medical devices.¹³⁰ Chemicalgenetic resources in fungal pathogens can aid in elucidating novel virulence pathways that can be targeted by small molecules, as well as identify the precise virulence factor(s) that are targeted by such compounds. For example, one such study employed the C. albicans GRACE collection to identify 102 negative morphogenetic regulators and 872 positive regulators, implicating ergosterol biosynthesis and N-linked glycosylation in morphogenesis.94 Another study using the GRACE library implicated the Arp2/3 complex in C. albicans adherence and biofilm formation, due to its role in modulating cell surface hydrophobicity and remodeling of chitin and β -glucans in the fungal cell wall.¹³¹ Finally, an approach combining genome-wide C. albicans transcriptional analysis and mining S. cerevisiae chemical genomic data successfully pinpointed inhibition of the mitochondrial retrograde response via MGE1 as the probable mechanism of niclosamide, a repurposed antihelmintic agent that disrupts *C. albicans* filamentation and biofilm formation.¹³²

Concluding thoughts: steady steps forward on the long road ahead

This review has highlighted the tremendous progress that has been made over the past few decades in the development of functional genomic resources that have advanced our understanding of how small molecules impact the fungal cell. While several hurdles remain to translate what we as a community have learned into development of the next antifungal drug that can be deployed in the clinic, the diverse array of cellular targets identified provides hope that there remains a plethora of cellular targets to be exploited to combat mycotic disease with much needed single agent and combination therapy treatments. The continued expansion of chemical genetic resources in diverse fungal pathogens will greatly empower inference of mode of action and resistance mechanisms for novel candidate antifungals. Likewise, elucidation of pathogen-specific drug-target pairs will further empower structure-activity relationship analysis and optimization of more species-selective and effective analogs. Beyond human health applications, chemical genetic approaches are also being implemented directly in fungal plant pathogens for agricultural fungicide development, which is similarly facing unprecedented resistance pressures.¹³³ Overall, although the generation of functional genomic resources in diverse fungal species requires substantial investment of time and resources, it is exquisitely clear that these advances will catalyze the development of much-needed antifungal drugs.

Acknowledgments

We thank all members of the Cowen lab for helpful discussions. L.E.C. is supported by the Canadian Institutes of Health Research Foundation Grant (FDN-154288), two National Institutes of Health NIAID R01 Grants (1R01AI127375-01 and 1R01AI120958 – 01A1), and a National Institutes of Health NIAID R21 Grant (1R21AI141080-01). L.E.C. is a Canada Research Chair (Tier 1) in Microbial Genomics & Infectious Disease and Co-Director of the CIFAR Fungal Kingdom: Threats & Opportunities program.

Competing interests

L.E.C. is a cofounder and shareholder in Bright Angel Therapeutics, a platform company for development of novel antifungal therapeutics; and a consultant for Boragen, a small molecule development company focused on leveraging the unique chemical properties of boron chemistry for crop protection and animal health.

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