Modular development enables rapid design of media for alternative hosts

Andrew M. Biedermann[1,2], Isabella R. Gengaro[1,2], Sergio A. Rodriguez-Aponte[1,2], Kerry R. Love[1,2], J. Christopher Love[1,2]

[1]Department of Chemical Engineering, Massachusetts Institute of Technology, Cambridge, Massachusetts 02139, United States
[2]The Koch Institute for Integrative Cancer Research, Massachusetts Institute of Technology, Cambridge, Massachusetts 01239, United States

*Correspondence to: clove@mit.edu

Target journal: Biotechnology & Bioengineering
Article (<15 double-spaced pages of text and no more than 10 figures/tables)

Keywords: Media development; media blending; *Pichia pastoris*; alternative hosts; automation

Abstract

Developing media to sustain cell growth and production is an essential and ongoing activity in bioprocess development. Modifications to media can often address host or product-specific challenges, such as low productivity or poor product quality. For other applications, systematic design of new media can facilitate the adoption of new industrially relevant alternative hosts. Despite manifold existing methods, common approaches for optimization often remain time and labor intensive. We present here a novel approach to conventional media blending that leverages stable, simple, concentrated stock solutions to enable rapid improvement of measurable phenotypes of interest. We applied this modular methodology to generate high-performing media for two phenotypes of interest: biomass accumulation and heterologous protein production, using high-throughput, milliliter-scale batch fermentations of *Pichia pastoris* as a model system. In addition to these examples, we also created a flexible open-source package for modular blending automation on a low-cost liquid handling system to facilitate wide use of this method. Our modular blending method enables rapid, flexible media development, requiring minimal labor investment and prior knowledge of the host organism, and should enable developing improved media for other hosts and phenotypes of interest.

Introduction

Achieving high volumetric productivities of biologic drugs in cultivation is a key step in

advancing candidate biologic drugs. The outcome of this effort ultimately impacts

manufacturing costs as well as readiness for transitioning clinical-stage development (Love,

Love, & Barone, 2012). The development of standard, chemically defined media for established

manufacturing hosts, such as CHO, has made such transitions efficient for monoclonal

antibodies by achieving high biomass accumulation, cell viability, operational consistency, and

specific productivities, streamlining development efforts (McGillicuddy, Floris, Albrecht, &

Bones, 2018; Rodrigues, Costa, Henriques, Azeredo, & Oliveira, 2012). Nonetheless, optimizing

productivity or quality attributes for a specific product often still requires further refinement of

media (Ritacco, Wu, & Khetan, 2018). Such development may require evaluating dozens of

variants derived from a common standard formulation to address the specific challenges

encountered (Gagnon et al., 2011; Loebrich et al., 2019). Media development for entirely new

biomanufacturing technologies, such as alternative hosts (Matthews, Kuo, Love, & Love, 2017a)

or new product modalities (Lu et al., 2016), may also require new formulations or extensive

optimizations due to limited prior knowledge.

Common approaches to develop a medium to optimize a phenotype of interest are often

labor intensive, low throughput, or rely heavily on extensive analytical capacity (Galbraith,

Bhatia, Liu, & Yoon, 2018). For example, analysis of residual media after cultivation requires

extensive capabilities for analytical characterization and prior experience with the manufacturing

host to identify potentially limiting or toxic media components (Mohmad-Saberi et al., 2013;

Pereira, Kildegaard, & Andersen, 2018). As a result, optimizations can be slow and iterative.

Furthermore, for an alternative host such as *Komagataella phaffii* (formerly known as *Pichia*

pastoris), there is substantially less, if any, prior knowledge available to establish profiles for

residual components in media after fermentation. Other analytical techniques like RNA-seq

combined with methods for reporter metabolite analysis can guide media optimization, to

generate testable hypotheses regarding beneficial modifications to media (Matthews et al.,

2017a). Such genome-scale approaches, however, require prior host-specific knowledge, such

as well-annotated genomes, and are still limited by slow iteration and labor-intensive

preparations of new media to test the hypotheses generated from computational analyses.

Alternative strategies for blending basal components for media allow linear combinations

of existing media to explore many variations rapidly (Jordan, Voisard, Berthoud, & Tercier,

2013). While this approach avoids slow iterative analyses, the typical experiment is labor

intensive to perform, often requiring independent preparations of over a dozen stock media to

combine (Rouiller et al., 2013). Similar to analytical-based approaches for optimization, the

selected variations of media are simultaneously guided and constrained by prior experience and

media designs, which may limit the breadth of components examined (Kennedy & Krouse,

1999). For less established hosts with fewer available formulations of media, media blending

may also require fully *de novo* formulations for initial studies. Further complicating such designs,

different and new components for media can present challenges in solubility or unanticipated

interactions with other elements in the formulations (Ritacco et al., 2018). New approaches to

blending could, however, enable fast, flexible experimentation and minimize the time, labor, and

analytical development needed initially to optimize media for new applications and phenotypes.

Here, we present a novel and generalizable approach for the modular development of

media and demonstrate its use to create optimized media for two different phenotypes—cellular

growth and recombinant expression of a protein (as measured by the secreted heterologous

protein titer) from *Pichia pastoris*. Our approach comprises two modular parts for blending and

optimization. We determined that a set of simple concentrated stock solutions constructed in

defined modules could generate many media by blending or dilution. We then automated a

simple, inexpensive liquid handling system (Opentrons OT-2) to enable high-throughput

screening for the effects of diverse media on a phenotype of interest in milliliter-scale batch

cultures. To maximize the benefit of this automated blending, we also developed an algorithmic

framework for systematic modular media optimization, beginning from a simple minimal media

(here a YNB-based one). This framework provides insights pertaining to key media components

during stages of optimization, as well as overall mapping of the design space for the media. In

the examples presented here, the resulting defined media developed with this strategy

outperformed commonly used BMGY and BMMY complex formulations for biomass

accumulation and secreted heterologous protein production.

Materials and Methods

Strains and cultivation conditions

Media for evaluating biomass accumulation were developed using a previously

described strain expressing G-CSF under control of the pAOX1 promoter (Crowell et al., 2020).

24-well plate screens were conducted as described previously, except cells were grown on a

labForce shaker and were only sampled 24 hours after inoculation (Matthews et al., 2017a).

BMGY, BMMY, and RDM media were formulated as described previously for shake flasks

(Matthews et al., 2017a). All cultivations were inoculated from a working cell bank at an initial

cell density of 0.1 OD/mL. For each working cell bank, cells grown in 1 L shake flasks with a 200

mL working volume of RDM were harvested during exponential growth (4-5.5 OD/mL) via

centrifugation at 1500 rcf for 4 minutes at 23 $^{\circ}$C and resuspended in an equal volume mixture of

RDM and 50 v/v% glycerol. This mixture was then distributed into 700 μL aliquots and stored at

-80 $^{\circ}$C, resulting in a cell density of ~30 OD/mL for the cell bank.

Media for evaluating enhanced production were developed using a strain expressing a

rotavirus-derived subunit vaccine candidate, P[8], under the control of the pAOX1 promoter

described previously (Dalvie et al., 2020). Biomass accumulation proceeded for 24 hours; cells

reached an initial induction density of ~20 OD/mL. Cultures were then exchanged into

production media and allowed to produce protein for an additional 24 hours. Supernatant was

harvested by centrifugation at 1500 rcf for 4 minutes at 23 $^{\circ}$C and filtered using a Captiva 96

well 0.2 μm filter plate (Agilent Technologies, Santa Clara, CA) prior to titer measurement by

RP-UHPLC.

Media components and supplements were purchased from Sigma-Aldrich, St. Louis,

MO, unless otherwise indicated in the supporting information. A table of supplement and stock

solutions with screening concentrations is also included in the supporting information. During

modular optimization, all media were prepared in high throughput using an Opentrons OT-2

liquid handler (Opentrons, Brooklyn, NY, software version $\geq$3.16.1) using Openblend. Modular

media blending code and instructions for setup and operation are provided in the Openblend

package (https://github.mit.edu/lovelab/openblend). For consistency, media used in final head-

to-head comparisons were prepared in bulk and filter sterilized through a 0.2 μm benchtop filter.

Analytical procedures

Biomass was measured by optical density at 600 nm as described previously (Matthews

et al., 2017a). An Agilent Bravo liquid handler was used to dilute samples prior to

measurements of OD into the Tecan Infinite M200 Pro plate reader.

Reverse phase ultra-high performance liquid chromatography (UHPLC) analysis was

performed on Agilent 1290 Infinity II UHPLC system controlled using OpenLab CDS software

(Agilent Technologies, Santa Clara, CA). The concentration of protein was determined using a

Poroshell 120 SB-Aq column (2.1 x 50 mm, 1.9µm) operated at 1.0 mL/min and 70 $^{\circ}$C (Agilent

Technologies, Santa Clara, CA). Buffer A was 0.1% (v/v) TFA in water and buffer B was 0.1%

(v/v) TFA, 0.5% (v/v) water in ACN. A gradient was performed as follows: 30% B for 1 min., 30-

40% B over 3 min., 40-90% B over 0.5 min., 90% B for 0.5 min., 90-30%B over 0.5 min., and

30% for 1 min.; total method run time was 6.5 minutes. Sample injection volumes were 50µL. A

diode array detector was set for absorbance detection at 214nm. Data analysis was completed

using OpenLab CDS Data Analysis (Agilent Technologies, Santa Clara, CA).

Statistical analysis and DOE design was conducted using JMP (SAS Institute, Cary NC).

Quadratic models were fitted using effect screening and non-significant terms (adjusted p-value

> 0.01) were eliminated sequentially in order of decreasing adjusted p-value to avoid overfitting.

Data was plotted using Prism 8.4.0 (GraphPad Software, San Diego, CA).

Results

Design of approach for modular media blending

We sought to develop an approach capable of identifying important, beneficial

modifications for media tailored to a given phenotype of interest. We reasoned that key

requirements for such an approach would be that it is fast and automatable, with minimal

dependence on complex analytical assay development. Such features would enable routine

application to any measurable phenotype of interest. In general, media blending allows both

speed and low analytical complexity. We aimed to retain these features while minimizing the

labor and constraints on compositions imposed by linear combinations of fully formed and

unique media. We reasoned that diverse and flexible blends of media could be created by

defining simple concentrated stock solutions as basic modules to combine further. These

modules would comprise individual components or common subsets of components with

compatible solubilities (e.g. YNB). If media components could be formulated in concentrated

stock solutions that could be stored stably over time, then the components could be routinely

and interchangeably combined and diluted to the desired final concentrations. This approach

would yield a broadly applicable modular strategy for media blending amenable to conventional

liquid handling automation.

To test the feasibility of this approach, we first assessed whether many common media

components could be formulated in concentrated stable aqueous stock solutions. Using the

CHO medium eRDF as a reference, we estimated the solubility of each component of this

medium, using data from AqSolDB as well as other online sources (Combs, 2012; FSA Panel

on Additives and Products or Substances used in Animal Feed (FEEDAP), 2011; Ritacco et al.,

2018; Schnellbaecher, Binder, Bellmaine, & Zimmer, 2019; Sorkun, Khetan, & Er, 2019;

Yamamoto & Ishihara, n.d.). We compared the estimated solubility of each media component to

its concentration in eRDF and found that, individually, most media components are soluble at

levels >10x higher than their eRDF concentration (**Figure 1A**). The existence of a wide range of

commercially available concentrated supplements further supports this result: >50x

concentrated solutions of amino acids, vitamins, lipids, and trace metal supplements are

common and commercially available.

Next, we used the product information of commercially available supplements, literature

sources, and inspection to estimate the percentage of eRDF media components that could be

stored in stable solutions for >6 months. We estimated that over 75% of eRDF components met

this criterion (**Figure 1B**). To address stability challenges caused by less stable components,

we reasoned that less stable components or supplements, such as vitamins, could be prepared,

aliquoted, and stored frozen for long-term storage (Schnellbaecher et al., 2019); these aliquots

could then be thawed and used within a defined period to mitigate component instability and

enable their integration into our modular blending strategy. Together, these solubility and

stability data suggested that a modular approach to media development could be defined in this

way to accommodate a range of new formulations easily.

We next automated the process for constructing media, using the Opentrons OT-2. We

chose this liquid handler due to its low cost, reliability, and compatibility with simple formats for

data input, such as Excel spreadsheets. We then created an open-source Python package,

named Openblend, which simplified the media construction process by handling routine

experimental design and execution steps (Figure 1C). Openblend creates an experimental

design spreadsheet, specifying the number of 24 well plates, the desired media composition of

each well, and stock solution names and concentrations. The script then checks the feasibility

of the experimental design, ensuring that the total volume of each well will not exceed the target

volume and avoiding the addition of sub-microliter stock solution volumes. If the design passes

this assessment, the script then outputs a new spreadsheet containing the setup for the OT-2

deck and required volumes of stock solutions, providing a user with instructions on how to setup

the OT-2 liquid handler. We found that our typical time to execute this script, setup the OT-2

and initiate plate building was ~15 minutes, and the time for the automated steps was about two

hours.

Finally, we defined a modular approach for optimization to effectively leverage the

Openblend tool (**Figure 1D**). Beginning from an initial basal medium, improved media are

constructed through successive rounds of optimization. In each round, a library of media

components and supplements are screened to identify beneficial additives. These additives are

then screened in combination and over a range of concentrations to further optimize the

performance of the medium. Each modular addition and optimization of additives can be guided

simply by measurements of the phenotype of interest (e.g. biomass accumulation). This greedy

approach to multi-dimensional optimization could continue iteratively until the resulting media

met desired specifications, all available media components were explored, or no additional

gains in performance realized.

Application to Developing a Medium for Biomass Accumulation

To assess the utility of this blending-based approach, we next aimed to identify and

optimize the concentration of media components beneficial for rapid biomass accumulation of *P.*

pastoris in batch cultivation. We previously described a rich defined medium (RDM) (Matthews

et al., 2017a), capable of high growth rates during biomass accumulation. One challenge

encountered with this formulation, however, was that precipitates can form at higher pH values

that require filtering during bulk preparations. Nonetheless, this medium provided a relevant

comparison for assessing the medium realized with our new approach due to its prior

demonstrated benefits relative to complex media. Following our modular approach, we

improved biomass accumulation by optimizing the accumulated optical density at 600 nm after

24 hours of cultivation.

234   Algorithms for optimizing systems based on multiple dimensions are often sensitive to

initial conditions used (Zakharova & Minashina, 2015). Given this potential confounding effect

here, we tested first the effects of the types of carbon source, nitrogen source, and pH set point

on biomass accumulation, using 1x YNB without amino acids or ammonium sulfate (YNB) to

satisfy minimum requirements for the concentrations of trace elements. We conducted a full-

factorial DOE using glycerol, glucose, and fructose as carbon sources; urea and ammonium

sulfate as nitrogen sources; and potassium phosphate as a buffer with pH values of 5, 5.75, and

6.5. We selected initial concentrations of 40 g/L, 4 g/L urea or the N-mol equivalent for

ammonium sulfate, and 10 g/L potassium phosphate, similar to values used in other media for

Pichia pastoris (Matthews et al., 2017a). A least squares regression model, including individual,

combination, and quadratic effects was fit to the log of optical density after 24 hours, a proxy

variable for the average growth rate (R^2 = 0.81). We determined that the two most significant

model terms were the type of carbon source and the interaction of the nitrogen source with pH

(**Figure 2A**). We found that cells grew significantly faster on metabolically related sugars

(glucose and fructose) than on the polyol (glycerol) commonly used for Pichia during biomass

accumulation (**Figure 2B**). This result affirms prior reports where glucose has been used for

biomass accumulation of Pichia (Guo et al., 2012; Moser et al., 2017).

251   The model also suggested that poor biomass accumulation during cultivation resulted

from a combination of ammonium sulfate as a source of nitrogen with low buffer pH (**Figure

2B**). This outcome may result from the production of acidic species associated with cellular

ammonium metabolism in the batch cultivation (Villadsen, 2015). Interestingly, the model

indicated slightly greater biomass was achieved with urea instead of ammonium sulfate. The

biomass accumulation of cultures grown with urea as a source of nitrogen were less sensitive to

reduced pH values (~5). We observed, however, that cultivations at pH 5 showed extensive

flocculation compared to those at 6.5. Given the insensitivity of urea-fed cultivations to buffer

pH and the high solubility and potential for low-cost sourcing of fructose, we therefore chose to

include fructose, urea, and a potassium phosphate buffer with a pH of 6.5 in our initial media

formulation.

With this basal formulation determined, we next screened for concentration-dependent

interactions of other key additives to the media and then optimized concentration-dependent

parameters. Following the same approach for screening effects, we conducted a full factorial

DOE over a broad range of media component concentrations: YNB (0.5, 1, 2x), fructose (10, 30,

50 g/L), urea (1, 4, 7 g/L), and potassium phosphate adjusted to a pH of 6.5 (4, 10, 16 g/L). The

resulting model identified fructose as a concentration-sensitive parameter (R^2=0.73) (**Figure**

2D). Terms involving the concentration of YNB were also highly ranked, but not statistically

significant. No significant interactions between components were identified in the model. We

therefore sought to better understand the concentration dependence of fructose and YNB

independently (**Figure 2E**), over an 8-fold range of concentrations. As expected, biomass

accumulation was highly sensitive to fructose concentration, with an optimum around 22.5 g/L of

fructose. The concentration of YNB had minimal effect on biomass accumulation; the presence

of trace elements supplied by YNB, however, was essential to growth. Based on these results,

we chose concentrations of 22.5 g/L fructose, 1x YNB, 7 g/L urea, and 10 g/L potassium

phosphate buffer. We reasoned that although biomass accumulation was relatively insensitive

to the concentrations of YNB and urea, higher concentrations could provide improved media

depth in future applications. We named this basal formulation DM1_dev0.

We next assessed what additional media components could improve biomass

accumulation. To test over 60 different components individually would require over 60 individual

solutions. Such an approach would scale linearly with new components; instead, we chose to

screen groups of related components, using commercially available pre-mixed supplements. We

compiled a library of 16 commercial supplements and industrially-relevant surfactants containing

more than 60 unique components and screened their individual effect on biomass accumulation

after 24 hours. In this way, we reasoned we could efficiently identify critical classes of

components related to the phenotype of interest and potentially deconvolve specific individual

additives of interest by inference. We used the recommended concentrations of each

supplement as supplied in product information, or critical micelle concentrations, and prior

knowledge for broad classes in yeast media to set reasonable screening concentrations

(Supporting Information). We identified five beneficial and two detrimental supplements that

significantly impacted biomass accumulation ($p_{adj} < 0.02$; 1-way-ANOVA) (**Figure 2F**). In

general, the results suggest that supplementation with amino acids and trace metals were

beneficial for accumulating biomass, while two surfactants, Tween 20 and CHAPS, were

detrimental. For this phenotype, the effects of vitamin and lipid supplements were minor;

supplements from either supplement category were not significantly beneficial or detrimental to

biomass accumulation. Our earlier experiments suggest that vitamins are essential but

concentration agnostic (**Figure 1E**), while lipid supplementation provides no clear benefit for

biomass accumulation.

Based on these results, we chose to test whether combinations of supplements of amino

acids and trace salts could yield synergistic improvements in biomass accumulation. We

screened pairwise combinations of the five beneficial supplements of mixed composition and

ranked the performance of our supplementation strategies (**Figure 2G**). A combination of 1x

MEM amino acids with 0.1 v/v% PTM1 salts resulted in the highest yield of biomass, though we

observed strong performance from other combinations of amino acid and trace metal

supplements. Based on these data, we chose to add MEM amino acids and PTM1 salts in our

basal medium and optimized their concentrations (**Figure 2H**).

Based on these results, we elected 0.1 v/v% PTM1 salts and 1x MEM amino acids, in

order to balance the moderate benefits and potentially high costs of amino acids. We found,

however, that the inclusion of the PTM1 salts in liter-scale preparations produced fine

precipitates, which can impede sterile transfers in use. To overcome this challenge, we

screened a broad range of PTM1 salts concentrations to identify the minimum concentration

required for improved outgrowth performance (**Figure 2I**). We found that PTM1 addition at

concentrations as low as 0.0005 v/v% led to increased biomass accumulation. We therefore

revised our PTM1 salts concentration to 0.01 v/v%, a concentration high enough to obtain the

benefits of PTM1 supplementation without inducing precipitate formation. This formulation we

named DM1.

Completing this series of optimizations with our iterative modular approach to define a

new formulation of medium, we then compared with other common media used to grow *P.*

pastoris. We evaluated the performance of this new optimized medium (DM1) relative to the

unsupplemented basal medium (DM1_dev0), the rich defined medium (RDM) we had previously

developed, and a common medium 4 v/v% glycerol BMGY. We found that DM1 yielded the

highest biomass accumulation, with significantly higher biomass accumulation relative to RDM

and BMGY (**Figure 2J**). This result demonstrates the utility of our modular strategy here for

media development that yielded an improved formulation for biomass accumulation compared

to other common media with minimal time and labor investment, and without requiring complex

analytical methods like mass spectrometry or RNA-sequencing.

Identifying media conditions important to heterologous protein production in K. phaffii

In addition to the time and labor savings of modular media development, our proof-of-

concept experiments demonstrated that this approach creates a flexible medium that can be

rapidly adapted to new growth phenotypes, as well as a data package that the identifies media

conditions important to the phenotype of interest. We reasoned that these additional benefits

could be particularly relevant for optimizing production of heterologous proteins. Understanding

which media components contribute most significantly to productivity could improve culture

performance and help identify important metabolic pathways or physiological effects for further

study.

To develop a medium for improved production of a recombinant protein, we chose to use

a strain engineered to secrete a rotavirus-derived subunit vaccine component, VP4-P[8], as a

model protein. We have previously demonstrated that this viral antigen can be expressed at

high titer under the control of the methanol-inducible pAOX1 promoter in BMMY media (Dalvie

et al., 2020). Similar to our initial approach to optimize a medium for growing biomass, we first

determined and optimized the concentrations of the sources for carbon and nitrogen, along with

the pH. The expression of P[8] in the strain tested uses the methanol-dependent pAOX1

promoter for inducible expression, so we selected methanol as the initial carbon source. We

then examined the impact of the source of nitrogen and buffer pH on titer. We conducted a full-

factorial DOE using identical concentrations as those used to create a medium for accumulating

biomass. The resulting model was visualized by ranking combinations of sources of nitrogen

and buffer (**Figure 3A**). The effects showed no interaction between these two factors. Urea was

again identified as the preferred source of nitrogen while higher pH values led to improved

secreted P[8] productivity. Unlike biomass accumulation, this pH dependence was observed

across both nitrogen sources.

We next applied the same DOE to identify important concentration-dependent

interactions that impact the production of P[8]. Unsurprisingly, the concentration of methanol

was the most important factor, with possible minor effects from other components (**Figure 3B**).

We decided to screen further a 20-fold range in methanol concentrations using two formulations

for remaining media components—the one determined for optimal cell growth (DM1) and the

optimal base media formulation predicted by the quadratic model here (2x YNB, 1 g/L urea, 4

g/L potassium phosphate adjusted to a pH of 6.5). We found that production was relatively

insensitive for concentrations of methanol ranging from 1-4 v/v%, with an optimum around 2%

(**Figure 3C**). We postulated that the rapid decline in productivity observed in these milliliter-

scale cultures using concentrations >6 v/v% methanol was likely due to excess formation of

toxic metabolic byproducts such as formaldehyde and hydrogen peroxide (Wakayama et al.,

2016). Interestingly, the predicted optimal medium from this set of studies outperformed the

medium we determined for accumulating biomass, suggesting that certain components of the

basal medium may benefit protein expression more than cellular growth and underscores the

value of optimizing media for specific phenotypes of interest. Based on these data in total, we

defined a basal medium for production including 2x YNB, 2 v/v% methanol, 1 g/L urea, and 4

g/L potassium phosphate buffer adjusted to a pH of 6.5 (DM2_dev0).

Next, we examined which supplements could improve the performance of DM2_dev0.

We added three chemical chaperones (TUDCA, sodium deoxycholate monohydrate (SDM), and

valproic acid) (Kuryatov, Mukherjee, & Lindstrom, 2013; Uppala, Gani, & Ramaiah, 2017), two

antioxidants (reduced glutathione (GSH) and N-acetyl cysteine (NAC)), and the chelator, K-

ETDA, to the list of 16 supplements included in our original screen defined for biomass

accumulation. Concentrations for these components were chosen based on product

specifications, literature data, and prior experience (Supporting Information). Many of the 22

supplements screened improved production of P[8] (**Figure 3D**). The top four ranking

supplements comprised surfactants or lipids, which could modulate membrane fluidity and lipid

metabolism (Butler, Huzel, Barnab, Gray, & Bajno, 1999; Degreif, Cucu, Budin, Thiel, & Bertl,

2019; Ritacco, Frank V; Yongqi Wu, 2018).

We then screened combinations of lipid supplements and surfactants to identify potential

synergistic effects. We ranked the individual supplements and their combinations (**Figure 3E**)

according to the measured titers of P[8]. We found that the addition of a cholesterol-rich

supplement yielded the highest secreted titers of P[8] (~50% improvement compared with

supplement-free condition in initial screens). Interestingly, a synthetic cholesterol supplement

alone did not substantially improve performance, suggesting the benefit results from a

combination of fatty acids and surfactant components in the supplement (**Supporting**

Information). This conclusion is consistent with similar improvements observed from other

supplements, such as linoleic acid-oleic acid-albumin (**Figure 3D**).

Since no other synergistic effects were observed in the combination screen, we

assessed the dependence of titer on the concentration of the cholesterol-containing supplement

identified (**Figure 3F**). Similar to our observations with cellular YNB used in the outgrowth

media, we found that concentrations of the supplement as low as 0.2 v/v% were beneficial for

protein expression, but that production was relatively insensitive to concentration (**Figures 3F,**

3G). We then directly compared the supplemented medium to the original composition; the new

supplemented media provided a 25% improvement in titer (p = 0.0006, one-tailed Welch's T

test). This new formulation with 1x cholesterol supplement, which we named DM2_dev1, was

the result of one cycle of optimization using our method.

Components of the cholesterol supplement included fatty acids, cholesterol, and

cyclodextrin, which are all are known to modulate membrane fluidity, a key parameter in vesicle

trafficking (Cooper, 1978; Degreif et al., 2019; Mahammad & Parmryd, 2015). We reasoned that

the addition of this supplement could therefore have synergistic effects with other supplements,

but did not find any further supplementation that improved P[8] titers within our original screen

(**Figure 3H**). We, therefore, considered if there could be additional classes of beneficial

supplements, absent from the original screen. Previous experiments demonstrated that P[8]

productivity is highly sensitive to methanol concentration (**Figure 3C**), so we wondered whether

further modulation of central carbon metabolism could yield additional productivity gains.

Modification of central carbon metabolism is best accomplished by feeding cells

alternative carbon sources, either entirely or as co-feeding substrates. Four co-fed substrates

have previously been shown to be non-repressive of pAOX1: sorbitol, mannitol, trehalose, and

alanine (Inan & Meagher, 2001). These substrates can be co-utilized with methanol without

repressing the pAOX1 promoter, which controls expression of P[8]. We hypothesized that the

introduction of supplemental carbon sources could enable further optimization of central carbon

metabolism. We screened co-fed substrates individually and in 1:1 combinations at a total

concentration of 20 g/L (a concentration similar to the optimal fructose and methanol

concentrations observed in previous carbon source optimizations) (**Figure 2E,3C**). Sorbitol co-

feeding had the most beneficial effect, resulting in a ~80% increase in P[8] titer (**Figure 3I**).

Mannitol supplementation was also beneficial (~70% increase), while alanine and trehalose co-

feeding were detrimental to productivity. While co-feeding carbon sources led to increased

biomass yield during production, these differences did not account for the improved titer, as

improvements in specific productivity (q_p) of ~60% and ~45% were also observed for the sorbitol

and mannitol co-fed conditions, respectively (**Supporting Information**). Based on these data,

we chose to include sorbitol as a supplemental carbon source for further study.

The addition of a supplemental carbon source could significantly impact central carbon

metabolism. We, therefore, wondered how the inclusion of sorbitol might impact the optimal

carbon feeding strategy. Examining total carbon source concentrations from 20 – 70 g/L, we

compared the performance of cultures co-fed with sorbitol:methanol ratios of 3:1, 1:1, and 1:3 to

a methanol-only control (**Figure 3J**). All co-fed conditions outperformed the methanol-only

control, suggesting that the presence of sorbitol is highly beneficial for producing P[8]. The titer

was relatively insensitive to sorbitol:methanol ratios and carbon concentrations. Based on the

data, we decided to use 2 v/v% methanol and 20 g/L of sorbitol for the final sorbitol-

supplemented media named DM2.

Finally, we compared the P[8] titer obtained using DM2_dev0, DM2_dev1, and DM2 to

other common production media for *P. pastoris*: BMMY and RDM. We found that DM2 led to a

~2x improvement in P[8] titers, relative to BMMY and RDM, up to 97 $\pm$ 2 mg/L.

Discussion

Here we have implemented a novel and broadly applicable approach for media

development that relies on rapid, automated construction of diverse media from defined

modules of components. We demonstrated the utility of this approach by developing two new

media for two phenotypes of interest in the heterologous production of proteins by yeast,

namely biomass accumulation and secreted production. We systematically identified and

optimized the concentration of media components important to each phenotype of interest.

Importantly, defining these new formulations of media did not require advanced analytical

capabilities and required minimal experimental time to assess more than 360 total formulations

during two to three rounds of optimization for each.

Our optimized formulations affirmed the importance of lipid-related components for

maximizing titers in *Pichia pastoris* cultivations. The importance of optimizing membrane fluidity

or lipid metabolism has been well established in CHO and appears to be key to optimizing

heterologous protein secretion in *P. pastoris* cultivation as well (Clincke et al., n.d.; Ritacco et

al., 2018; Zhang, Wang, & Liu, 2013).

Modular media blending has four advantages over existing methods. First, the use of

common stock solutions and supplements to formulate media reduces initial labor required for

new experiments or optimizations ~15 minutes per experiment, making parallel testing of

multiple hypotheses efficient and requires less resources overall. Here, we created 30 stock

solutions, and evaluated >360 unique media compositions, without manual preparation of

individual media or extensive blending calculations or planning. Most of these solutions could be

readily reused in future experiments to optimize for new phenotypes of interest. Second, our

method requires minimal knowledge of the host organism *a priori* and could, in principle, be

applied to any measurable phenotype of interest. We anticipate that this method could be used

to optimize other phenotypes of interest, such as glycosylation profiles. Third, our method

provides certain practical advantages, including minimal requirements for analytical

characterization and rapid identification of component interactions that lead to solubility

challenges. These traits make it possible to learn about formulations that may lead to extensive

precipitates like those encountered with our rich defined medium formulation (**Figure 4A**).

Finally, modularly constructed media, such as DM2, can be ~70% pure water with low

osmolarity, leaving volumetric and osmotic space for future modifications to accommodate new

or related phenotypes of interest (**Figure 4B**).

We also acknowledge certain limitations in the present study that may be addressed in

future work. First, while modular media development identifies components key to the

optimization of the phenotype of interest, additional media optimization effort may be necessary

to translate these learning in batch cultivations to scaled-up fed-batch or perfusion operation,

where additional variables such as supplemental feed composition and feeding schedule must

also be considered. In principle, modular media construction could be applied to high-

throughput scale-down cultivation models, such as Ambr250s. Second, our approach for

optimization relies on greedy algorithms tailored to create a new media for a single phenotype of

interest; however, given the vast explorable media space it is possible to find a local optimum.

Further metabolic or -omic modeling techniques could be employed to guide broader exploration

of media space, co-optimize multiple phenotypes, or facilitate biologically informed optimization,

albeit with more complex experimental and computational requirements (Matthews, Kuo, Love,

& Love, 2017b; Mohmad-Saberi et al., 2013). Third, our current method used commercially

available supplements, but in practice, beneficial supplements could be simplified by using

individual components, to facilitate more biological inferences and aid development of improved

host-specific supplements. Finally, initial screens to identify beneficial supplements rely on

reasonable choices of initial concentrations for screening. These currently require prior

knowledge from the literature or commercial sources; with further use in the community of the

Openblend approach, it is possible additional sharing of knowledge could help inform further

developments.

The improved speed and accessibility of in-depth media development experiments

enabled by modular media construction could help improve expression of many classes of

proteins in laboratories and discovery centers that have not traditionally had access to such

capabilities. Since many lead candidates for new therapeutic proteins begin in small biotech

firms and academic labs, early-stage improvements in productivity could help advance more

proteins towards the clinic simply by facilitating access to larger quantities of proteins for initial

research and non-clinical studies. In more established companies, the ability to make rapid

improvements to existing media may enable faster product development timelines and could

reduce manufacturing costs overall. Rapid identification and optimization of sensitive media

components could also enable easier adoption of a range of industrially relevant alternative

hosts, resulting in further manufacturing flexibility and potentially cost savings (Coleman, 2020).

Acknowledgements

The authors acknowledge Danielle Camp for program coordination. This work was

funded by the Bill & Melinda Gates Foundation (Investment ID INV-002740). The content is

solely the responsibility of the authors and does not necessarily represent the official views of

the Bill & Melinda Gates Foundation.

A.M.B., I.R.G., K.R.L., and J.C.L. conceived and planned experiments. A.M.B.

conducted media development experiments. S.R.A. developed and maintained the RP-UPLC

assay. A.M.B. performed analytical characterization. A.M.B., K.R.L., and J.C.L. wrote the

manuscript. J.C.L. and K.R.L. designed the experimental strategy and supervised analysis. All

authors reviewed the manuscript.

References

Butler, M., Huzel, N., Barnab, N., Gray, T., & Bajno, L. (1999). Linoleic acid improves the robustness of cells in agitated cultures, 27–36.

Clincke, M.-F., Guedon, E., Yen, F. T., Ea, L., Universite, N., Ogier, V., … Goergen, J.-L. (n.d.). Effect of Surfactant Pluronic F-68 on CHO Cell Growth , Metabolism , Production , and Glycosylation of Human Recombinant IFN- c in Mild Operating Conditions, 181–190. https://doi.org/10.1002/btpr.503

Coleman, E. (2020). Establishment of a Novel Pichia Pastoris Host Production Platform by.

Combs, G. F. (2012). *The vitamins: Fundamental aspects in nutrition and health*. San Diego: Elsevier Academic Press.

Cooper, R. A. (1978). Influence of Increased Membrane Cholesterol on Membrane Fluidity and Cell Function in Human Red Blood Cells. *Journal of Supramolecular Structure*.

Crowell, L. E., Crowell, L. E., Raymond, A., St, H. E., Engineering, C., Doyle, P. S., & Crowell, L. E. (2020). Accelerating process development for biologics on an automated , pharmacy-scale manufacturing system by by.

Dalvie, N. C., Brady, J. R., Crowell, L. E., Tracey, M. K., Biedermann, A. M., Kaur, K., … Love, J. C. (2020). Molecular engineering improves antigen quality and enables integrated manufacturing of a trivalent subunit vaccine candidate for rotavirus. *BioRx*, 1–51.

Degreif, D., Cucu, B., Budin, I., Thiel, G., & Bertl, A. (2019). Lipid determinants of endocytosis and exocytosis in budding yeast. *BBA - Molecular and Cell Biology of Lipids*, *1864*(7), 1005–1016. https://doi.org/10.1016/j.bbalip.2019.03.007

FSA Panel on Additives and Products or Substances used in Animal Feed (FEEDAP). (2011). Scientific Opinion on safety and efficacy of choline chloride as a feed additive for all animal species. *EFSA Journal*, *9*(9), 2353.

Gagnon, M., Hiller, G., Luan, Y. T., Kittredge, A., Defelice, J., & Drapeau, D. (2011). High-End pH-controlled delivery of glucose effectively suppresses lactate accumulation in CHO Fed-batch cultures. *Biotechnology and Bioengineering*, *108*(6), 1328–1337. https://doi.org/10.1002/bit.23072

Galbraith, S. C., Bhatia, H., Liu, H., & Yoon, S. (2018). Media formulation optimization: current and future opportunities. *Current Opinion in Chemical Engineering*, *22*, 42–47. https://doi.org/10.1016/j.coche.2018.08.004

Guo, C., Huang, Y., Zheng, H., Tang, L., He, J., Xiang, L., … Jiang, H. (2012). Secretion and activity of antimicrobial peptide cecropin D expressed in Pichia pastoris. *Experimental and Therapeutic Medicine*, 1063–1068. https://doi.org/10.3892/etm.2012.719

Inan, M., & Meagher, M. M. (2001). Non-repressing carbon sources for alcohol oxidase (AOX1) Promoter of Pichia pastoris, *92*(6), 585–589.

Jordan, M., Voisard, D., Berthoud, A., & Tercier, L. (2013). Cell culture medium improvement by rigorous shuffling of components using media blending, 31–40. https://doi.org/10.1007/s10616-012-9462-1

Kennedy, M., & Krouse, D. (1999). Strategies for improving fermentation medium performance: a review. *Journal of Industrial Microbiology and Biotechnology*, (23), 456–475.

Kuryatov, A., Mukherjee, J., & Lindstrom, J. (2013). Chemical Chaperones Exceed the Chaperone Effects of RIC-3 in Promoting Assembly of Functional a 7 AChRs. *PLoS ONE*, *8*(4), 1–11. https://doi.org/10.1371/journal.pone.0062246

Loebrich, S., Clark, E., Ladd, K., Takahashi, S., Brousseau, A., Kitchener, S., … Ryll, T. (2019). Comprehensive manipulation of glycosylation profiles across development scales. *MAbs*, *11*(2), 335–349. https://doi.org/10.1080/19420862.2018.1527665

Love, J. C., Love, K. R., & Barone, P. W. (2012). Enabling global access to high-quality biopharmaceuticals. *Current Opinion in Chemical Engineering*, *2*(4), 383–390. https://doi.org/10.1016/j.coche.2013.09.002

Lu, T. L., Pugach, O., Somerville, R., Rosenberg, S. A., Kochenderfer, J. N., Better, M., &
Feldman, S. A. (2016). A Rapid Cell Expansion Process for Production of Engineered
Autologous CAR-T Cell Therapies. *Human Gene Therapy Methods*, *27*(6), 209–219.
https://doi.org/10.1089/hgtb.2016.120
Mahammad, S., & Parmryd, I. (2015). Cholesterol depletion using methyl-β-cyclodextrin. In
Methods in membrane lipids (pp. 91–102).
Matthews, C. B., Kuo, A., Love, K. R., & Love, J. C. (2017a). Development of a general defined
medium for Pichia pastoris, (July), 103–113. https://doi.org/10.1002/bit.26440
Matthews, C. B., Kuo, A., Love, K. R., & Love, J. C. (2017b). Development of a general defined
medium for Pichia pastoris, (July), 103–113. https://doi.org/10.1002/bit.26440
McGillicuddy, N., Floris, P., Albrecht, S., & Bones, J. (2018). Examining the sources of variability
in cell culture media used for biopharmaceutical production. *Biotechnology Letters*, *40*(1),
5–21. https://doi.org/10.1007/s10529-017-2437-8
Mohmad-Saberi, S. E., Hashim, Y. Z. H. Y., Mel, M., Amid, A., Ahmad-Raus, R., & Packeer-
Mohamed, V. (2013). Metabolomics profiling of extracellular metabolites in CHO-K1 cells
cultured in different types of growth media. *Cytotechnology*, *65*(4), 577–586.
https://doi.org/10.1007/s10616-012-9508-4
Moser, J. W., Prielhofer, R., Gerner, S. M., Graf, A. B., Wilson, I. B. H., Mattanovich, D., &
Dragosits, M. (2017). Implications of evolutionary engineering for growth and recombinant
protein production in methanol - based growth media in the yeast Pichia pastoris. *Microbial*
Cell Factories, 1–16. https://doi.org/10.1186/s12934-017-0661-5
Pereira, S., Kildegaard, H. F., & Andersen, M. R. (2018). Impact of CHO Metabolism on Cell
Growth and Protein Production: An Overview of Toxic and Inhibiting Metabolites and
Nutrients. *Biotechnology Journal*, *13*(3), 1–13. https://doi.org/10.1002/biot.201700499
Ritacco, Frank V; Yongqi Wu, A. K. (2018). Cell Culture Media for Recombinant Protein
Expression in Chinese Hamster Ovary (CHO) Cells : History , Key Components , and
Optimization Strategies. https://doi.org/10.1002/btpr.2706
Ritacco, F. V., Wu, Y., & Khetan, A. (2018). Cell culture media for recombinant protein
expression in Chinese hamster ovary (CHO) cells: History, key components, and
optimization strategies. *Biotechnology Progress*, *34*(6), 1407–1426.
https://doi.org/10.1002/btpr.2706
Rodrigues, M. E., Costa, A. R., Henriques, M., Azeredo, J., & Oliveira, R. (2012). Comparison of
commercial serum-free media for CHO-K1 cell growth and monoclonal antibody
production. *International Journal of Pharmaceutics*, *437*(1–2), 303–305.
https://doi.org/10.1016/j.ijpharm.2012.08.002
Rouiller, Y., Périlleux, A., Collet, N., Jordan, M., Stettler, M., & Broly, H. (2013). A high-
throughput media design approach for high performance mammalian fed-batch cultures,
(June), 501–511.
Schnellbaecher, A., Binder, D., Bellmaine, S., & Zimmer, A. (2019). Vitamins in cell culture
media: Stability and stabilization strategies. *Biotechnology and Bioengineering*, *116*(6),
1537–1555. https://doi.org/10.1002/bit.26942
Sorkun, M. C., Khetan, A., & Er, S. (2019). AqSolDB , a curated reference set of aqueous
solubility and 2D descriptors for a diverse set of compounds, *2019*(July), 1–8.
https://doi.org/10.1038/s41597-019-0151-1
Uppala, J. K., Gani, A. R., & Ramaiah, K. V. A. (2017). Chemical chaperone , TUDCA unlike
PBA , mitigates protein aggregation efficiently and resists ER and non-ER stress induced
HepG2 cell death. *Scientific*, *1*(January), 1–13. https://doi.org/10.1038/s41598-017-03940-
1
Villadsen, J. (2015). Redox Balances and Consistency Check of Experiments. In *Fundamental*
Bioengineering (pp. 17–38).
Wakayama, K., Yamaguchi, S., Takeuchi, A., Mizumura, T., Ozawa, S., Tomizuka, N., …

Nakagawa, T. (2016). Regulation of intracellular formaldehyde toxicity during methanol
metabolism of the methylotrophic yeast Pichia methanolica. *Journal of Bioscience and*
Bioengineering, *122*(5), 545–549. https://doi.org/10.1016/j.jbiosc.2016.03.022
Yamamoto, T., & Ishihara, K. (n.d.). Stability of Glutathione in Solution. *Developments in Food*
Engineering, 209–211.
Zakharova, E. M., & Minashina, I. K. (2015). Review of Multidimensional Optimization Methods.
Mathematical Models, Computational Methods, *60*(6), 625–636.
https://doi.org/10.1134/S1064226915060194
Zhang, H., Wang, H., & Liu, M. (2013). Rational development of a serum-free medium and fed-
batch process for a GS-CHO cell line expressing recombinant antibody, 363–378.
https://doi.org/10.1007/s10616-012-9488-4
625
626

Figure 1. Modular media development can be broadly applicable, easily applied, and

systematically executed to improve measurable phenotypes of interest.

A) Estimate of the ratios of component solubility to their concentrations in medium demonstrates

that most components are soluble at >10x their concentration in the CHO medium, eRDF. B)

With the exception of some classes of medium components, such as vitamins, most media

components can be formulated into solutions that remain stable for >6 months under proper

storage conditions. C) Overview of time, labor and planning saved by using Openblend to

automate modular media construction. D) Overview of a modular media optimization approach,

636 which can be used to build an optimized medium for any measurable phenotype of interest

637 systematically.

639
Figure 2. Modular development of a new biomass accumulation media for *P. pastoris*

A) Significance of carbon (fructose, glucose, glycerol), nitrogen (urea and ammonium sulfate),

and pH choice (5, 5.75, 6.5) in a least square regression model fitted to a full factorial DOE. B)

Fructose and glucose were found to result in significantly higher biomass accumulation after 24

hours of outgrowth than glycerol. C) Ammonium sulfate was found to be more pH sensitive than

urea, as shown by the JMP sensitivity profiles during fructose feeding. D) Significance of terms

in a least square regression model fitted to a full factorial DOE over fructose, urea, potassium

phosphate, and YNB concentrations. E) 1-FAAT optimization of fructose and YNB concentration

finds optimal outgrowth performance at a fructose concentration of 22.5 g/L and relative

insensitivity over a wide range of YNB concentrations (0.15 to 1.2x). F) A media

supplementation screen identified 5 beneficial supplements, related to trace element and amino

acid supplementation. G) Further screening of beneficial supplement combinations identified

synergistic amino acid and trace metal supplementation strategies. H) Comparison of the effect

of MEM amino acid concentration on biomass accumulation at different PTM1 salts

concentrations. I) Effect of the concentration of PTM1 salts on biomass accumulation in

DM1_dev0 medium supplemented with 1x MEM AA. J) Head-to-head comparison of 4 v/v%

glycerol BMGY, 4 v/v% glycerol rich defined medium, the initial defined biomass accumulation

media (DM1_dev0), and the final biomass accumulation medium obtained after a full

optimization cycle (DM1), demonstrates that DM1 leads to superior biomass accumulation.

661
Figure 3. Modular development of a media for heterologous protein production in *P. pastoris*

A) Initial full-factorial screen of nitrogen source choice and buffer pH demonstrates that urea is

preferred over ammonium sulfates and high buffer pH is preferred over lower values. B) A full-

factorial concentration optimization identified methanol as the most concentration dependent

variable. Other components in the base media were predicted to affect productivity with much

lower levels of significance. C) Evaluation of the effect of methanol concentration on P[8] titer,

using two different base media (urea, buffer, and YNB concentrations): the biomass

accumulation base medium and the optimal base media composition predicted by our

concentration DOE. D) Ranking of supplements according to their effect on P[8] titer.

Supplements related to membrane fluidity or lipid metabolism ranked highly. E) Evaluation of

combinations of lipid and surfactant supplements confirmed that cholesterol supplementation

leads to the greatest improvement in P[8] titer. F) Concentration optimization of cholesterol

demonstrated low concentration dependence, with similar performance observed over a 40-fold

range (0.2-8x). G) Comparing cholesterol-free and cholesterol-supplemented cultures fed at

various concentrations demonstrates that cholesterol supplementation results in a significant

~25% improvement in P[8] titers ($p<0.001$). H) No significantly beneficial supplements were

observed when repeating the supplementation screen. I) Screening supplementation of 20 g/L

of co-fed substrates individually or in 1:1 combinations by mass identified sorbitol

supplementation as highly beneficial to P[8] titer. J) Examination of the effect of co-feed ratio

and total carbon concentration on titer in DM2_dev1 supplemented media. K) Comparison of

P[8] titer obtained with DM2 to previous iterations and other common *P. pastoris* media

demonstrates a ~2x improvement in P[8] titer, relative to 1 v/v% methanol RDM and 1 v/v%

methanol BMMY.

686

Figure 4. Comparison of DM2 to rich define medium.

A) Comparison of precipitate formation during construction of RDM (left) and DM2 (right) media.

Adjusting the pH of RDM to 6.5 results in significant formation of white precipitate. No

precipitate formation is observed in DM2. B) Relative volumes of stock solutions and pure water

needed to construct DM2. Pure water addition accounts for 69% of DM2 volume, demonstrating

that there is substantial room for further supplement exploration and development. When

separated into simple stock solutions, DM2 can be 3x concentrated.