The Cost of Protein Production

Graphical Abstract

What limits protein production?

DNA

RNA

Protein

Low Phosphate:

Transcription limitation

Low Nitrogen:

Translation limitation

Highlights

- Libraries expressing increasingly high protein amounts are extensively studied
- Processes that limit protein production vary, depending on growth conditions
- Ribosomes are not universally limiting in rapidly growing cells
- Cells adapt by increasing their size and the abundance of endogenous proteins

Authors

Moshe Kafri, Eyal Metzl-Raz, Ghil Jona, Naama Barkai

Correspondence

naama.barkai@weizmann.ac.il

In Brief

Kafri et al. investigate the processes that limit protein production. They find that enforcing either gene transcription or protein translation reduces growth rate, depending on growth conditions. Cells adapt by increasing their size and endogenous proteome content, suggesting that rapidly growing cells are not resource limited.
The Cost of Protein Production

Moshe Kafri,1,3 Eyal Metzl-Raz,1,3 Ghil Jona,2 and Naama Barkai1,*
1Department of Molecular Genetics, Weizmann Institute of Science, Rehovot 76100, Israel
2Department of Biological Services, Weizmann Institute of Science, Rehovot 76100, Israel
3Co-first author
*Correspondence: naama.barkai@weizmann.ac.il
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SUMMARY

The economy of protein production is central to cell physiology, being intimately linked with cell division rate and cell size. Attempts to model cellular physiology are limited by the scarcity of experimental data defining the molecular processes limiting protein expression. Here, we distinguish the relative contribution of gene transcription and protein translation to the slower proliferation of budding yeast producing excess levels of unneeded proteins. In contrast to widely held assumptions, rapidly growing cells are not universally limited by ribosome content. Rather, transcription dominates cost under some conditions (e.g., low phosphate), translation in others (e.g., low nitrogen), and both in other conditions (e.g., rich media). Furthermore, cells adapted to enforced protein production by becoming larger and increasing their endogenous protein levels, suggesting limited competition for common resources. We propose that rapidly growing cells do not exhaust their resources to maximize growth but maintain sufficient reserves to accommodate changing requirements.

INTRODUCTION

Protein production in living cells is tightly coordinated with external conditions and intracellular demands. This regulation ensures that needed proteins are produced, while those whose function is not compatible with current needs are not. It may also serve to minimize the cost of protein production; indeed, making proteins consumes cellular resources by using energy and nutrients as building blocks and also by occupying common cellular machineries such as ribosomes, polymerases, or chaperones, whose abundance may be limiting. Defining the cost of protein production, and the molecular processes limiting this production, is a central challenge in cellular physiology, which is relevant in particular for understanding the interplays among cell growth, cell division, and cell size.

The cost of protein production is commonly attributed to protein translation (Andrews and Hegeman, 1976; Emilsson and Kurland, 1990; Kurland, 1992; Marr, 1991; Molin et al., 1974; Scott and Hwa, 2011; Scott et al., 2010; Vind et al., 1993). A major fraction of the cellular GTP pool is used for amino acid polymerization, while significantly lower quantities are invested in other processes including gene transcription and protein folding (Russell and Cook, 1995; Schimmel, 1993). Ribosomes were implicated as the major factor limiting growth of rapidly growing cells (Dennis et al., 2004; Emilsson and Kurland, 1990; Klumpp et al., 2013; Maaloe and Kjeldgaard, 1966; Marr, 1991; Russell and Cook, 1995; Scott et al., 2010, 2014; Vind et al., 1993), following the discovery that most of the cellular biosynthetic activity is devoted to making ribosomes (Bremer and Dennis, 1996) and the observation that ribosome content is tightly coordinated with cell growth rate (Bremer and Dennis, 1996; Schaechter et al., 1958; Warner, 1999). However, there is still a need for experimental data that directly map the molecular mechanisms limiting protein expression.

The cost of protein production was extensively studied in E. coli. Forced overexpression of the Lac operon in medium lacking lactose leads to reduced cell growth and arrested the cell cycle when reaching ~30% of total proteome (Dong et al., 1995; Horiuchi et al., 1962; Nguyen et al., 1989; Novick and Weiner, 1957; Scott et al., 2010). This reduced growth was broadly interpreted as the cost of protein production (Dong et al., 1995; Klumpp et al., 2013; Scott et al., 2010; Stoebel et al., 2008; Zamenhof and Eichhorn, 1967), although a recent study provided an alternative interpretation (Eames and Kotermame, 2012). Only limited data are available describing protein burden in eukaryotic cells (Hauf et al., 2000; Lang et al., 2009; MacLean, 2007), and it is not clear whether results inferred from one cell type, or from specific conditions, can be generalized to other organisms and environments.

In this paper, we examined the relative contributions of gene transcription and protein translation to the cost of protein production in budding yeast, S. cerevisiae, and mapped the limitation to the initiation versus elongation steps of each process. We found that transcription and translation can both be limiting, depending on the growth conditions. In particular, our data challenge the hypothesis that ribosome content is a universal limiting factor defining growth rate of rapidly growing cells, as only slow-growing cells appeared to be limited in ribosome content. Notably, endogenous protein expression was increased, rather than decreased, upon forced production of inert proteins, suggesting that the protein production capacity can readily adapt to increasing demands. We discuss the implications of our data for describing the interplay between protein production and cell growth.
RESULTS

Systematic Study of Protein Burden

Microorganisms are optimized for rapid growth. We therefore analyzed the impact of enforcing protein production on cell growth rate. Expressing mCherry fluorescent protein using one of the strongest yeast promoters (pTDH3, glyceraldehyde-3-phosphate dehydrogenase enzyme) resulted in a change in growth rate that was below our experimental resolution (~2%).

To enforce higher expression levels, we cloned the pTDH3-driven mCherry construct into a plasmid that integrates into the genome in multiple copies and generated a library of strains expressing one up to ~20 copies of this construct (Figures 1A, 1B, and S1A), as measured by mCherry fluorescence and verified by qPCR of genomic DNA (Figure 1C). Proportionality between fluorescence levels and DNA copy numbers was maintained for the full range of integrations and was compatible with protein abundances measured by western blots (Figures 1D and S1B). A comparable library was generated also using GFP, giving consistent results (Figure S3B). We calibrated the measured fluorescence levels using 15 GFP-fused proteins and available abundances data (Figures 1E and S1C; Table S3; Supplemental Experimental Procedures). By this, we estimate that a single pTDH3-GFP construct contributes ~2% to the cellular proteome.

Growth Fitness Decreases in Proportion to the Expressed Proteins

We measured the growth rate of our library strains using two methods. First, we used direct comparison of growth curves of strain expressing a single copy or 18 integrations of pTDH3-mCherry. The strain with 18 copies grew ~18% slower than the strain expressing a single copy (Figure 1F). Second, relative growth rates were quantified for the full library using a sensitive competition assay (Figure 1G). Strains containing different integration number of pTDH3-mCherry were co-incubated with GFP-labeled wild-type cells, and their relative abundance in the population of growing cells was quantified at subsequent generations.

Figure 1. Generating Libraries of Strains Expressing Increasing Amounts of Chromosomally Integrated Fluorescence Proteins

(A) pTDH3 promoter driving mCherry expression (pTDH3-mCherry) was cloned into a plasmid that, when transformed into yeast cells, integrates into the genome in multiple copies (Figure S1A). (B) mCherry expression range can be appreciated from the differences in the colonies’ color (left panel, natural light). Fluorescence images of individual cells expressing low or high mCherry copies (middle panel) and superimposed with bright light (right panel). No aggregates were observed in high-burdened cells (n > 1,000 cells). Scale bar, 5 μm. (C) Differences between colonies were quantified using a flow cytometer (fluorescence-activated cell sorting [FACS]) measuring mCherry fluorescence and by qPCR. The FACS copy number was obtained by normalizing using the 1 copy strain fluorescence. Error bars indicate SE of three biological repeats. (D) Quantification of the mCherry proteins using western blots. Error bars indicate SE. (E) Calibration curve defining the fraction of cellular proteins made by a single pTDH3-GFP construct. Fifteen distinct GFP-fused strains spanning a range of abundances were chosen. Their fluorescence levels were quantified using a flow cytometer and plotted as a function of their protein abundance in PPM (parts per million) as described in Supplemental Experimental Procedures and Figure S1C. Horizontal error bars represent SE between different measurements in Table S3, and vertical error bars represent the SD between different fluorescence measurements. Vertical dashed blue line indicates 2.3% from the proteome.
times using a flow cytometer. When cells were grown in rich media (YPD) and maintained in log phase by serial dilutions, the relative growth rate decreased linearly with increasing mCherry fluorescence (Figure 1H). Also by this estimate, each additional copy of the construct reduces the growth rate by $\sim$1% (Figure S1D).

To examine whether protein burden depends on growth conditions, we repeated our measurements when growing cells in different media, this time allowing cells to enter and exit stationary phase. In all conditions tested, growth rate decreased linearly with increasing mCherry expression, although to different extents (Figures 2A, 2C, and S2). Notably, cells growing in low-phosphate conditions appeared less sensitive to the introduced burden compared to cells growing in standard media, possibly reflecting their somewhat lower division rate. Exit from prolonged starvation was also delayed (Figure 2B), as was reported (Shachrai et al., 2010). Finally, when growing cells in continuous cultures, burdened cells reached a lower steady-state density (Figure 2D), indicating that they require a higher influx of the limiting nutrient for maintaining the same imposed growth rate.

Repeating our experiment using an additional library of lower-expression-capacity strains did not show measurable growth defects, verifying that the reduced fitness is not due to the multiple genomic integrations (Figures 3A, S3C, and S3D). In contrast, replacing wild-type GFP by its $\sim$20-fold less stable variant (Mateus and Avery, 2000) had the same effect on fitness, confirming that growth is not impaired by accumulating GFP (Figures 3B and S3A). Consistent with this, neither the mCherry nor the GFP proteins formed detectable aggregates (Figures 1B and S3F–S3H; Supplemental Information). Because Tdh3p is a key enzyme in glycolysis (McAlister and Holland, 1985), the reduced growth rate of our strain could reflect impaired glycolysis resulting from competition with endogenous TDH3 expression. Fluorescence levels in our library, however, increased linearly with the DNA copy number (Figure 1C), suggesting limited competition between copies. Furthermore, two additional libraries, in which mCherry was driven by different promoters ($pPDC1$ and $pPGK1$), showed practically the same fitness cost when normalized by expression levels (Figure 3C). More directly, we verified that the expression of Tdh3-GFP-fused protein was not lower in burdened cells in any of the conditions tested (Figure S3I). Therefore, we can conclude that the reduced fitness of our library strains does not result from the multiple genomic integrations, protein toxicity, or impaired glycolysis, suggesting that it is the result of a limitation in the protein production process itself.

Figure 2. Fitness Cost of Forcing Unneeded Protein Expression Is Condition Dependent

(A) The fitness cost of protein expression depends on the growth conditions. The reduced growth rate of pTDH3-mCherry expressing cells was quantified in the different specified conditions and is plotted as a function of the mCherry fluorescence. Note that in all cases, the growth rates are reduced in proportion to mCherry expression, but the magnitude of the fitness cost depends on the tested condition. The cells were grown on SC (blue), low-phosphate (0.2 mM phosphate, red), and low-nitrogen media (50 μM of ammonium sulfate, cyan). For more conditions, see Figure S2A. (B) Protein burden during recovery from starvation. Cells were grown in SC to stationary phase and were left in this media for 1 week before being transferred back to fresh media. Shown is the recovery time (hours) from starvation, defined by the time required for 50% increase in cell density. (C) Generation time in the different media tested. Error bars indicate SE of three biological repeats. See also Figures S2C–S2H. (D) The OD$_{600}$ in steady state in continuous culture (dilution time: SC, 3 hr; low phosphate and low nitrogen, 4 hr). Measurements’ mean during 5–12 hr with SD lower than 0.005 between the different measurements.

Gene Transcription and Protein Translation Both Contribute to Protein Burden, Depending on Growth Conditions

We next wished to examine the relative contribution of protein translation and gene transcription to the cost of protein production. To this end, we generated an additional library in which mCherry was expressed from a destabilized mRNA (DAmP [decreased abundance by mRNA perturbation]; Figure 4A) (Muhlrad and Parker, 1999; Schuldiner et al., 2005), thereby reducing the translation load while maintaining the same transcription burden. Indeed, when expressed using the same TDH3 promoter, the DAmP strain produced $\sim$10-fold lower
mCherry fluorescence (Figure S4A) and ~30-fold less mRNA as measured by qPCR (not shown). The burden on the translation machinery was thus reduced by ~10-fold. By comparing fitness cost in these two libraries, we could therefore distinguish the cost coming from transcription from that coming from the translation process; a cost that results from burdened translation depends only on the amount of mCherry produced, independently of whether it was transcribed from wild-type or destabilized mRNA. In contrast, the cost originating from limiting translation will depend on the genomic copy number (transcribed genes), irrespective of how many proteins were eventually produced (Figure S4B).

We applied this approach to cells growing at three different conditions: low phosphate, low nitrogen, and standard media (SC). The relative contribution of transcription versus translation to the measured burden varied depending on growth conditions. In medium low in phosphate, the cost scaled with the genomic copy number, indicating that transcription is limiting. In contrast, in medium low in nitrogen, the cost scaled with mCherry fluorescence, indicating that translation is limiting. In standard media (SC), both transcription and translation contributed about equally to the fitness cost (Figures 4B–4D, S4C, and S4D).

Cells growing in low phosphate are therefore limited in their transcription capacity, while cells growing in low nitrogen are limited in their translation capacity. To further verify that, we examined whether these different conditions inflict differential sensitivity to drugs that inhibit protein translation or gene transcription. Indeed, cells growing in low nitrogen were more sensitive to the translation-inhibiting drug hygromycin B, while cells growing in low phosphate were more sensitive to the transcription-inhibiting drug phenanthrolone (Figures S4E–S4I; Supplementary Information).

Initiation and Elongation Contribute to Protein Burden in a Condition-Dependent Manner

We next wished to map the measured limitations to the initiation or the elongation steps of the respective transcription/translation processes. To this end, we generated an additional library, in which mCherry ORF (open reading frame) was fused to GFP ORF. The two ORFs were separated by a linker, and were transcribed as a single unit using the TDH3 promoter (Figure 5A). mCherry fluorescence levels were invariant to the addition of GFP (Figure S5B). Comparing the fitness of the new library to the original mCherry library enabled us to distinguish the relative contribution of initiation and elongation to protein burden. In standard media (SC), the fitness cost scaled with mCherry fluorescence, irrespectively of GFP production, indicating that initiation is limiting. In contrast, in medium low in nitrogen, the cost scaled with the total length of the transcribed proteins, indicating that elongation is limiting (Figures 5B and 5D).

Introducing a stop codon between the mCherry and GFP, reduced mCherry levels, likely reflecting nonsense-mediated mRNA decay (Hentze and Kulozik, 1999; Figures 5C, S5B, and SSC). Fitness cost of expressing this construct in low-nitrogen medium still scaled with the total length of the translated proteins (mCherry only) (Figure 5D, left). Together, this maps the limitation in low nitrogen to translation elongation. In contrast, when growing this library in low phosphate, fitness cost scaled primarily with the genomic copy number and not with protein fluorescence, consistent with transcription initiation being the main limiting factor (Figure 5D, right). Fitness cost here was still somewhat lower than that of the fused Cherry-GFP, suggesting a minor contribution of translation elongation.
Translation Initiation Is Limiting in Slow-Growing Respiring Cells

Ribosome content and energy are critical resources for protein translation and are often implicated in limiting protein production (Andrews and Hegeman, 1976; Emilsson and Kurland, 1990; Klumpp et al., 2013; Maaløe and Kjeldgaard, 1966; Marr, 1991; Scott et al., 2010, 2014; Vind et al., 1993). We were therefore surprised that translation did not emerge as the universal, or major, factor limiting growth of our rapidly proliferating cells. Furthermore, even in the one case where limitation was mapped exclusively to translation elongation (low-nitrogen medium), it is more likely due to shortage of amino acids previously reported under these conditions (Boer et al., 2010). We reasoned that this might reflect the unconstrained availability of energy expected from fermenting cells, which direct the vast majority of their carbon molecules to produce ethanol (Figure 6A). Indeed, each glucose molecule generates two ATP molecules when fermented, but it can generate 36 ATPs if respired, so that a minor re-direction of the carbon flux toward respiration could provide all additional energy required when the translation capacity is forced to increase. As a result, we predicted that energy, and subsequently translation, would become limiting when growing cells on a sub-optimal carbon source that cannot be fermented.

To test this possibility, we grew our cells on a mixed ethanol-glycerol media. Cells division time was ~6 hr (Figure S6A), compared to 1.5 hr for cells grown on glucose-supported media. As predicted, limitation was mapped exclusively to translation; the reduced fitness of the DAmP and wild-type strains scaled with the amount of produced proteins, irrespective of mRNA stability (Figures 6B and S6B).

Energy limitation could directly impact on translation elongation and perhaps indirectly impact ribosomal machinery, as producing ribosomes is the major biosynthetic activity of the cell (Bremer and Dennis, 1996; Russell and Cook, 1995). We found that the cost of expressing the fused mCherry-GFP protein was the same as that of expressing mCherry alone, indicating that the limitation is due to translation initiation (Figures 6C, 6D, and S6C). While ribosome content could limit either initiation or elongation, a recent theoretical analysis concluded that initiation is the rate-limiting step in ribosome function (Shah et al., 2013). Together, our results suggest that during this slow growth on a non-fermentable carbon source, protein production is limited by ribosome content.

Cells Adapt to the Enforced Protein Production by Increasing Their Cell Size and Making More Endogenous Proteins

It is often assumed that microbes are optimized for rapid growth (Schaechter et al., 1958). Accordingly, it is expected that cells grow at their maximal possible capacity and cannot further increase their overall biosynthesis. In this framework, enforcing expression of unneeded proteins would necessarily come at the expense of producing other proteins (Klumpp et al., 2013; Scott et al., 2010; Shachrai et al., 2010). Our finding that ribosome content is not limiting under conditions of rapid growth, prompted us to re-visit this assumption. If protein production is not constrained, cells might grow larger when forced to express excessive amounts of proteins. We therefore examined whether
Figure 5. The Relative Contribution of Initiation and Elongation to the Fitness Cost of Protein Production

(A) Experimental strategy. The mCherry mRNA (black) was extended to also code for the GFP ORF (red). The two ORFs were separated by a linker, and an additional library was constructed. When driven by the same TDH3 promoter, cells expressing this allele showed the same mCherry fluorescence (Figure S5B). Therefore, the mCherry and mCherry-GFP libraries posed the same burden on the initiation process but differed in the burden on elongation.

(B) Distinguishing the contribution of initiation versus elongation to the fitness cost. Shown is the relative growth rate of strains in the specified libraries, which is plotted as a function of the mCherry fluorescence (left) and as a function of the length of the translated proteins (right). (Top) SC media. Note that fitness cost scales with the level of mCherry fluorescence, independently of the presence of GFP, indicating low contribution of elongation to the cost. (Middle) Limiting nitrogen. Note that fitness cost is dominated by elongation. (Bottom) Limiting phosphate. Note that both initiation and elongation contribute to the fitness cost.

(C) mCherry-Stop construct. As in (B), the mCherry mRNA was extended to code for GFP ORF. However, in this construct, the linker between the two ORFs contains translation stop codons. An additional library was constructed using this construct. The addition of the stop codon resulted in a lower mRNA levels, presumably due to nonsense mediated decay (Hentze and Kulozik, 1999; Figures S5B and S5C).

(D) Distinguishing between the contribution of transcription elongation versus translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost. Shown is the fitness cost of all three libraries specified in (A) and (C). (Left) Limiting nitrogen. All phenotypic cost in translation elongation to the fitness cost.
DISCUSSION

Common notion holds that microorganisms, including bacteria and yeast, are optimized for rapid growth, making the most efficient use of nutrients once these become available in order to outgrow other microorganisms that share the same habitat. Proteins are the growth driving force, but their production is a major consumer of energy and nutrients. What are the fundamental limits on cell growth rate? What cellular resources are limiting, and how are they distributed? What limits protein production, and how do these limitations impact the cell division cycle? Our study touches on these questions by systematically interrogating the origins and consequences of protein burden: the reduced growth rate of cells that are forced to express increasing amounts of endogenous proteins, suggesting a general adaptation of the protein production machinery and a limited competition for common resources.

Our results call for revisiting some common notions invoked for describing the physiology of cell growth. The first is the question of whether maximal growth rate is set by optimal allocation of ribosomes, dedicating the maximal possible fraction of ribosomes to the task of producing new ribosomes. This notion comes largely from the realization that cell growth is fundamentally limited by ribosome translation. For example, if 50% of the ribosomes are occupied in making new ribosomes, each one of those ribosomes will need to translate two ribosomes within each cell cycle. In E. coli, ribosome translation requires 6–10 min, in agreement with their 20-min minimal division time (Scott et al., 2010). Our results imply that budding yeast does not work in this limit, consistent with the larger difference between cell-cycle time (~90 min) and the time it takes to translate a eukaryotic ribosome (~24 min; Planta and Mager, 1998; Waldron et al., 1977; Supplemental Information). Thus, ribosome content was not universally limiting in the rapid growth conditions we tested. When we grew cells in medium containing low levels of phosphate (while maintaining rapid growth; Figure 2C), limitation was mapped to transcription initiation. Cells growing in SC were equally limited by transcription and translation initiation, whereas in yeast growing in low-nitrogen conditions, limitation was mapped to translation elongation (Figure 6D). This may be attributed to a shortage of ribosomes, but we find it more likely to result from the shortage of amino acids that has been described for this medium (Boer et al., 2010). Therefore, ribosome activity readily adapts to the increasing translation demands, at least in some conditions. A likely implication of our results is therefore that rapidly growing cells do not produce proteins at their maximal possible capacity. This, again, calls for revisiting the...
A common assumption that protein production is set by maximizing the use of some limiting resource, be it the ribosome, nutrient, or any other factor. Indeed, we show that cells readily adapt to the increased protein production by increasing in size and the amount of endogenous proteins. If ribosomes, or other factors, are not limiting protein production and cell growth, what defines a cell’s growth rate, and how does it depend on the number of proteins produced? Our data suggest that growth control involves the coordination of a variety of processes that function in parallel, with different growth conditions altering the relative importance of these processes. Proposed cell physiological models (Bremer and Dennis, 1996; Keren et al., 2013; Klump et al., 2013; Scott et al., 2010; Zaslaver et al., 2009) provide the necessary rigorous framework for formulating such limitations and testing possible hypotheses about their consequences. Still, the space of possible models is too broad considering the scarcity of available experimental data. The experimental setup we have described provides data essential for restricting and refining the resource allocation models toward the goal of understanding cell physiology and, in particular, the interplay among protein production, cell size, and cell growth.

Yeast, like bacteria, is a fast-growing organism. By contrast, in higher eukaryotes, most cells grow slower. Further, while budding yeast metabolize glucose by fermentation, respiration is predominant in most cells predominantly. By these criteria, these cells would be more similar to our glycerol-grown cells, in which ribosomes appear to be limiting. It would be interesting to extend similar studies for these different cell types.

**EXPERIMENTAL PROCEDURES**

**Strains and Media**

All of the budding yeast (S. cerevisiae) strains in this study were based on BY4741, BY4742 (Brachmann et al., 1998), or Y8205 (Tong and Boone, 2007) laboratory strains. Manipulations were performed using a standard PEG;LiAC;ssDNA protocol (Gietz and Woods, 2002). For strains and plasmids
used in this research, see Supplemental Experimental Procedures and Table S1; for strain stability, see Supplemental Information.

Strains were grown in YPD medium, YPEG medium, SC medium, or SC medium depleted of a specific nutrient, as described in the main text. Glucose-limiting media contained 0.2% glucose. For starvation experiments, cells were grown to stationary phase in SC medium overnight (optical density [OD] ~10) and were left without shaking at 30 °C for 1 week. See Supplemental Experimental Procedures for further details.

**Competition Assays**

Cells were grown to stationary phase in the relevant media overnight. GFP and mCherry strains were then co-incubated in the specified media at 30 °C. Frequencies of GFP versus mCherry cells were measured by flow cytometry. The cells were diluted once a day and may have reached stationary phase (except for the YPD log experiment, in which cells were diluted every ~9 hr). For more details and calculations, see Supplemental Experimental Procedures. The full list of competition experiments performed is given in Table S2.

**qPCR**

Cells were harvested and frozen immediately in liquid nitrogen; DNA and RNA were extracted using MasterPure Complete DNA and RNA Purification Kit (Epicenter, MC85200), respectively. Next, RNA was reverse-transcribed to cDNA using M-MLV Reverse Transcriptase kit (Promega, M1701). cDNA and DNA levels were compared to the wild-type (WT) using ACT1 gene as reference (LightCycler 480, Roche).

**Growth Curves and Generation Time**

Cells were grown overnight in the different media, then diluted to OD<sub>600</sub> ~0.1 into shake flasks. ODs were measured (Novaspec Plus, Amersham Biosciences) every 90/40 min (low-nitrogen medium/other media, respectively) for ~12 hr and a final OD<sub>600</sub> measurement after ~24 hr. Next, growth curves were plotted: log2(OD) as a function of time in hours. The generation time was calculated as 1/ s', where s' is the slope's linear part in the growth curve. 

**Quantitative Western Blots**

Post-alkaline NaOH protein extraction protocol was performed followed by standard quantitative western blot. See Supplemental Experimental Procedures for further details.

**Soluble, Insoluble, and Total Protein Isolation and Relative Quantification**

Soluble, insoluble, and total protein isolation and relative quantification were performed as previously described (Geier-Samerotte et al., 2011). Quantitative western blots were performed as described above.

**Microscopy**

Cells were grown overnight to stationary phase, then diluted to a final OD of ~0.5. The photos were taken by a Delta Vision 1 microscope (Applied Precision) with the following optics objective: Olympus 60×/1.42, Plan Apo N.

**Cell Size**

All the strains were diluted after overnight growth and left growing to OD<sub>600</sub> ~0.3. Next the cultures were diluted 1:40 with 0.5 M NaCl and immediately measured in a Multisizer4 Coulter counter (Beckman Coulter) (Coulter, 1953).

**GFP Level Measurements**

Haploids from the SGA selection, which possessed both the GFP-tagged protein and the different copies of mCherry, were grown overnight in the relevant media. The following day, cells from each plate were diluted 20 x and mixed 1:1.5 (low:high burden) and were left to grow with shaking at 30 °C for ~5 hr (~6 hr for low-nitrogen medium) before analysis in the flow cytometer (~80,000 cells per well). See Supplemental Experimental Procedures for further details.

**Data Analysis and Modeling**

For the protein abundance calculation of one genomic copy strain, see Supplemental Experimental Procedures.

For protein cost origin analysis and modeling and GFP haploid libraries analysis, see Supplemental Experimental Procedures.

**SUPPLEMENTAL INFORMATION**

Supplemental Information includes Supplemental Experimental Procedures, seven figures, and four tables and can be found with this article online at http://dx.doi.org/10.1016/j.celrep.2015.12.015.

**AUTHOR CONTRIBUTIONS**

M.K. and E.M.-R. designed experiments, performed research, interpreted data and wrote the manuscript. G.J. designed chemostat experiments and performed research, N.B. designed and supervised research, interpreted data, and wrote the manuscript.

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