Precise Post-translational Tuning Occurs for Most Protein Complex Components during Meiosis

Highlights

- The synthesis of most protein complex components during meiosis is imprecisely matched.
- The levels of most protein interaction partners are post-translationally adjusted.
- Ribosomal proteins are degraded and re-synthesized late in the meiotic program.
- Analysis of meiotic protein levels over time points to additional Ama1-APC/C targets.

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In Brief

Eisenberg et al. leverage global translation and protein data to identify cases of regulated protein degradation in meiosis. Analyses of temporal trends reveal that members of protein complexes can be synthesized at ideal stoichiometry but that they are usually made imprecisely and their levels adjusted by degradation.
Precise Post-translational Tuning Occurs for Most Protein Complex Components during Meiosis

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SUMMARY

Protein degradation is known to be a key component of expression regulation for individual genes, but its global impact on gene expression has been difficult to determine. We analyzed a parallel gene expression dataset of yeast meiotic differentiation, identifying instances of coordinated protein-level decreases to identify new cases of regulated meiotic protein degradation, including of ribosomes and targets of the meiosis-specific anaphase-promoting complex adaptor Ama1. Comparison of protein and translation measurements over time also revealed that, although meiotic cells are capable of synthesizing protein complex members at precisely matched levels, they typically do not. Instead, the members of most protein complexes are synthesized imprecisely, but their protein levels are matched, indicating that wild-type eukaryotic cells routinely use post-translational adjustment of protein complex partner levels to achieve proper stoichiometry. Outlier cases, in which specific complex components show divergent protein-level trends, suggest timed regulation of these complexes.

INTRODUCTION

The protein complement of a cell defines its structure and function and is determined by the relative rates of synthesis and degradation for each protein present. The mechanisms and specificity determinants of synthetic processes in gene expression, especially transcription, have been well studied. In addition, the basic classes of mechanisms by which proteins are degraded within cells, including through regulated ubiquitin-based protein turnover, have been defined. However, it remains difficult to systematically determine the impact of regulated protein degradation in an unperturbed system, even at steady state. Comprehensively assessing the timing and specificity of protein degradation mechanisms in the context of cellular differentiation is an even greater challenge, but also particularly critical, as the transitions between sequential cellular stages require waves of both synthesis of new proteins and removal of pre-existing proteins.

Early examples of regulated degradation were identified by single-gene analyses, such as the case of Cyclin during the cell cycle and meiosis (Evans et al., 1983). Cyclin protein synthesis was observed to be constitutive, but the protein level fluctuated, leaving regulated protein degradation as the remaining explanation for the protein expression pattern observed. These observations ultimately led to the discovery of the conserved anaphase-promoting complex/cyclosome (APC/C)-based specificity mechanism, which is responsible for a key event in cell division (Irniger et al., 1995; King et al., 1995; Sudakin et al., 1995). Here we apply a similar approach, based on examining genome-wide protein patterns during a natural process to identify cases of protein degradation more globally. We recently generated a complex dataset that enabled these analyses, and that includes deep and matched mRNA-, translation-, and protein-level measurements during the natural process of meiotic differentiation in budding yeast (Cheng et al., 2018). These analyses allowed the identification of genes that we propose to be targets of the Ama1-APC/C and revealed degradation and re-synthesis of ribosomal proteins following meiosis in maturing gametes.

We also found strong and widespread evidence for post-translational adjustment of the relative levels of protein complex components over the natural process of meiotic differentiation. While such regulation has been shown to occur in mutant conditions that create an imbalance among macromolecular complex components (Abovich et al., 1985; Dephoure et al., 2014; Gorenstein and Warner, 1977; Ishikawa et al., 2017; Lam et al., 2007; Papp et al., 2003; Sung et al., 2016a; Torres et al., 2010; Warner, 1977; Warner et al., 1985), the effect that we observe is widespread in wild-type cells subjected to no external perturbations. Our analyses show that although eukaryotic cells are capable of synthesizing binding partners with effectively perfectly matched levels, they typically synthesize them at only roughly similar levels and fine-tune their stoichiometry through protein degradation. Finally, we find that divergence of protein-level trends of specific protein complex members from their partners is suggestive of functional differences for these members, including regulatory roles.
RESULTS

Protein Degradation Inferred from Comparison of Protein and Translation Data

Although our comprehensive dataset of gene expression through meiotic differentiation (Figure 1A) was previously analyzed only for transcriptional and translational regulation (Cheng et al., 2018), we found that comparison of translation and protein patterns could also be used to infer post-translational regulation of gene expression that accurately captures expected regulation.

For example, the protein for cyclin-dependent kinase (CDK) Cdc28 is required through much of meiosis and was present for a period of at least 3 hr after translation ceased, suggesting protein stability during this time frame (Figure 1B). In contrast, protein levels of the synaptonemal complex (SC) component Zip1 declined in concert with a decrease in translation, which is consistent with the known active degradation of this protein in late prophase (Figure 1C; Sourirajan and Lichten, 2008). In an extreme case of such regulation, the CDK inhibitor Sic1 showed decreased protein levels, despite high ongoing translation during meiotic S phase and prophase, which is consistent with the critical known regulation of this protein (Figure 1D; Dirick et al., 1998).

Protein-level Co-clustering Identifies Additional Candidate Targets of Ama1

We reasoned that we should be able to identify cases of coordinated degradation by simply looking for groups of proteins that show similar timing and degree of protein-level decrease between sequential meiotic time points. To this end, we calculated the ratio of the protein level at each time point to that of the previous time point for every protein quantified and performed hierarchical clustering (Figure 1E). This analysis revealed a variety of discrete patterns of protein-level change over meiosis. Most prominent in the degree of change observed was a group of genes for which protein abundance increased prior to and early past the normal stage of degradation in Ama1-dependent degradation. We focused on two genes in the Cdc5/Ndt80/Ssp1 cluster (Figure 1E), RNA-binding proteins Pes4 and Mip6, based on their known meiotic function and our ability to construct C-terminally tagged versions of their encoded proteins that could be visualized by western blot. Because clustering was based on similar overall protein-level changes over time, gene groupings were likely to share both synthesis and degradation mechanisms. The former was already clear, as all genes in this group are predicted targets of Ndt80, which activates its own transcription along with a large set of other targets in late prophase to enable the meiotic divisions (Chu and Herskowitz, 1998). Matched mRNA sequencing (mRNA-seq) data were consistent with this regulation, showing overlapping patterns of mRNA accumulation for all five genes (Figure 2A). Translation patterns were also generally similar for these genes, which is consistent with no reported strong translational regulation for their mRNAs (Figure 2B). Protein accumulation patterns were similarly comparable for these genes, as expected, but the later decrease in protein levels was even more similar, with nearly identical downward slopes in late meiosis I (MI) (Figure 2C). The degree of similarity of these protein down-slopes mirrored that of their highly coordinated mRNA up-slopes and suggested that the degradation of these proteins may be mediated by the same temporally restricted mechanism (Figures 2A and 2C).

We observed the expected wild-type pattern of protein accumulation for Ndt80 and Ssp1, and the relative persistence of epitope-tagged protein in cells deleted for AMA1 (Figures 2D, 2E, S1A, S1B, S1F, S1G, S1K, and S1L). We found that Pes4 and Mip6 also showed the persistence of high protein levels past the normal stage of degradation in ama1 Δ cells, suggesting that their degradation is at least partially dependent on Ama1-APC/C in mid-meiosis, as predicted by cluster analysis (Figures 1E, 2F, 2G, S1H, and S1I). We confirmed that this effect could not be explained by generally delayed meiotic progression in ama1 Δ cells (Figures S1C, S1D, S1M, S1N) and further compared to a protein, Stu2, which has a similar overall protein pattern to the other candidates but was not in the same discrete subcluster as Cdc5, Ssp1, Ndt80, Pes4, and Mip6 (Figure 1E). We found no change in Stu2 protein persistence in ama1 Δ cells compared to wild-type, suggesting that the effect that we observe is specific (Figures 2H, S1E, S1J, and S1O). While Pes4 and Mip6 were not previously identified as targets of regulated degradation, these proteins have a temporally restricted role, mediating the translational repression of several Ndt80 targets prior to...
Figure 2. Protein Co-clustering Is Predictive of Shared Degradation Regulation by Ama1

(A) mRNA levels over time for known and predicted Ama1 targets. The green box highlights a period of matched mRNA induction timing, consistent with known transcriptional co-regulation.

(B) Translation levels over time for known and predicted Ama1 targets.

(legend continued on next page)
late meiosis II (MI; Jin et al., 2017). We conclude that the Ama1-APC/C has targets beyond those previously characterized and that its activity during the meiotic divisions may coordinate the destruction of diverse proteins that are important for MI but unnecessary during MII.

Substructure in Protein Clustering of Genes Co-regulated for Protein Synthesis Suggests Post-translational Adjustment Based on Physical Interactions

As the comparison of protein-level changes among genes that were co-regulated for synthesis was informative for identifying Ama1-dependent proteins, we wondered whether such analyses could more generally elucidate cases of regulated protein degradation. We turned to a group of genes that we previously identified as showing highly similar protein synthesis patterns (Brar et al., 2012) and that are enriched for roles in recombination and SC formation, two processes involved in the physical linkage of "mom" and "dad" chromosomes before MI segregation (Figure S2, left). For the complex components that we were able to quantify by mass spectrometry, protein abundance patterns showed greater diversity than translation level patterns. In addition, functional subgroups emerged from the protein measurements that had more similar patterns than the group as a whole (Figure S2, right), and protein abundance-level clusters correlate with known physical interactions. For example, recombination and SC factors, which interact directly or indirectly with chromatin, clustered separately from the genes that are not thought to be involved in these processes (Figure S2, bold versus plain text). Furthermore, proteins that are known structural components of the SC, which associate with chromatin (and each other) during meiosis, formed a subcluster in the protein abundance data that is distinct from more direct regulators of recombination (Figure S2, orange versus purple; reviewed in Cahoon and Hawley, 2016; Zickler and Kleckner, 2015). Finally, this grouping included three characterized heterodimers, and each pair clustered together in the protein but not the translation data (Figure S2, green; Humphryes et al., 2013; Pochart et al., 1997; Shinohara et al., 2008). Since protein abundance measurements integrate translation and protein degradation, we propose that the more precise matching of protein levels for interaction partners results from the degradation of uncomplexed subunits.

This conclusion is consistent with a predominant model for protein complex assembly and homeostasis that is based on mutant analyses. It has been widely reported that members of stable protein complexes are present with equivalent steady-state protein levels and that this balanced stoichiometry is important for cellular fitness (Burke et al., 1989; reviewed in Harper and Bennett, 2016; Veitia and Potier, 2015). Experiments in which one protein complex component was decreased in expression, for example, revealed resultant decreases in other complex components (Abovich et al., 1985; Stevens and Davis, 1998). Overexpression of a single protein complex component can result in decreased protein stability for this component, resulting from its proteasome-mediated degradation (Abovich et al., 1985; Ishikawa et al., 2017; Stevens and Davis, 1998; Sung et al., 2016a; Warner et al., 1985). Aneuploid cells carrying an extra chromosome showed expression from this chromosome that was proportionally increased at the protein level for most genes, with the exception of protein complex members, which were dampened at the protein level relative to expectations based on mRNA and translation measurements (Dephoure et al., 2014). In addition, a pulsed mass spectrometry study in mammalian cell lines revealed an initial rapid proteasome-dependent phase of degradation of a subset of proteins, which were enriched for protein complex membership, suggesting that the oversynthesis of some complex members leaves a subset of newly synthesized proteins uncomplexed and thus unstable (McShane et al., 2016).

While it is established that protein-level adjustment through degradation can occur in diverse mutant and perturbed conditions, its prevalence in wild-type cells under natural conditions has been difficult to assess. It is known that members of most protein complexes do not show highly correlated mRNA levels (Gandhi et al., 2011; Liu et al., 2009), and in isolated natural cases, it has been shown that protein complex components are synthesized out of stoichiometry and adjusted to similar levels by protein degradation (Blikstad et al., 1983; Lehner and Lodish, 1988; Mueller et al., 2015). It has also been reported, however, that wild-type cells tend to synthesize protein complex components in proportion to the stoichiometry seen in final complexes, perhaps precluding the need for regulation at the level of degradation (Li et al., 2014; Taggart and Li, 2018). Coherently reconciling results from individual gene studies and large-scale studies, as well as between regulation in wild-type and perturbed cellular conditions, has been challenging. Quantitatively comparing the levels of mRNA, translation, and protein in parallel for complex partners in wild-type cells should address this problem, but this requires comparing sequencing- and mass spectrometry-based measurements. These two fundamentally different types of measurements show different dynamic ranges of detection, and by traditional mass spectrometry, precise direct comparison of measurements between proteins with different physical properties is difficult.

Precisely Matched Complex Component Synthesis Is Possible but Not Pervasive during Meiotic Differentiation

We reasoned that our dataset provided a unique opportunity to assess the degree of post-translational adjustment of levels of stable protein complex members in wild-type cells because of several advantageous properties of the data. First, we measured mRNA, translation, and protein from matched extracts, allowing their direct comparison. Second, we measured nearly 80% of...
the annotated yeast proteome. Third, we made measurements for several sequential time points, which allowed analysis of trends and correction for differences in dynamic ranges of detection for mass spectrometry and sequencing data. As a simple first test, we examined the patterns of mRNA, translation, and protein for several heterodimer pairs. We found that in the case of TUB1/TUB2—the primary tubulin heterodimer genes—levels of mRNA, translation, and protein over time were remarkably well matched for both components (Figure 3A). The degree of similarity was striking given that these genes are encoded on separate chromosomes and show disparate promoter, 5' UTR, and open reading frame (ORF) sequences. Furthermore, TUB1 mRNA is subject to splicing, while TUB2 is not. This suggests that eukaryotic cells are capable of synthesizing complex components even when the genes encoding them have different cis-regulatory regions and sequences. However, in the cases of three other heterodimer pairs initially investigated in detail—Rbg1/Tma46, Pob3/Spt16, and Gtr1/Gtr2—we observed some similarity between mRNA and translation for the two genes, but much higher similarity at the protein level, suggesting that meiotic cells synthesize these protein partners somewhat imprecisely and that their levels are

Figure 3. Meiotic Cells Are Capable of Perfect Synthesis Matching of Heterodimer Partners, but It Is Uncommon
(A–F) Z-score plots show gene expression level trends of mRNA levels (left), translation levels (middle), and protein levels (right) over all time points for pairs of genes, including (A) heterodimer Tub1 and Tub2, (B) heterodimer Rbg1 and Tma46, (C) heterodimer Pob3 and Spt16, (D) heterodimer Gtr1 and Gtr2, (E) sequential enzymes involved in purine nucleotide biosynthesis Ade1 and Ade2, and (F) sequential enzymes involved in histidine biosynthesis His2 and His5. (G) Correlation coefficients for translation and protein between annotated heterodimer partners are shown at left. Yellow represents higher, blue represents lower. The same scaling is used at upper right to compare to a subset of sequential enzymes in biosynthetic pathways. Below right, a summary of trends for heterodimers and adjacent enzymes in biosynthetic pathways. Heterodimer partners show a greater protein than translation correlation, while sequential biosynthetic enzymes show the opposite trend. (H) Cumulative distribution plots for the translation and footprint correlations in (G). Translation correlations are indistinguishable for the heterodimers and biosynthetic pathway genes, but protein correlations are significantly higher for the heterodimers as assessed by the Kolomogorov-Smirnov (K-S) test. See also Figures S2, S3, and S4.
adjusted post-translationally to achieve 1:1 stoichiometry (Figures 3B–3D).

To determine the degree to which these examples reflect general trends, we isolated data for all of the genes noted as members of heterodimeric complexes in the Saccharomyces genome database (https://www.yeastgenome.org). We restricted our analysis to cases in which we were able to quantify both heterodimer members at all levels (mRNA, translation, and protein) for every time point and for which the heterodimer did not clearly involve prevalent partner substitution by either entirely different proteins or their paralogous partner. This last filter was a necessary simplification for some heterodimers, but the determination of “prevalence” was sometimes nuanced. For example, TUB3 was not included in our heterodimer analysis of TUB1/TUB2 because—although it is known that Tub3 can substitute for Tub1 in interacting with Tub2—Tub3 was synthesized at very low levels during meiosis, ranging from only 7% to 17% of those of Tub2. Our approach resulted in a list of 37 high-confidence heterodimer pairs (Figure 3G, left). The two heterodimers that make up the annotated tetramer for the Ndc80 kinetochore complex were also included (Spc24/Spc25 and Ndc80/Nuf2), because the temporal meiotic regulation of this complex (discussed below) has been studied in detail (Chen et al., 2017; Chia et al., 2017; Miller et al., 2012). Among these 39 cases, we found that 9 showed a pattern similar to that seen for Tub1/Tub2, with correlation coefficients for both translation and protein for the 2 genes that were higher than 0.95. Of the remaining 30 cases, 26 (87%) showed patterns suggestive of post-translational adjustment of protein levels, with protein-level correlations for the two genes exceeding translation-level correlations, by a large margin in some cases (Figures 3G, left, 3H, and S3C). This result suggests a model by which heterodimer partners are usually synthesized imprecisely and adjusted by degradation of unbound partners in unperturbed wild-type cells.

We considered the possibility that the higher protein compared to translation or mRNA correlations could simply reflect the differences in sequencing and mass spectrometry measurements, or the instantaneous nature of translation versus steady-state nature of protein abundance measurements. To determine whether protein measurements simply tended to correlate better than translation measurements for technical reasons, we examined gene pairs that should have similar expression patterns, but not because of stable physical association. We reasoned that sequential enzymes in linear biosynthetic pathways represent such a class of gene pairs and would be expected to show correlated expression, but not necessarily protein association. We again excluded cases in which paralogs serve major redundant roles and were again restricted to cases in which both sequential pathway components were quantified in our dataset at all levels. We included 21 protein pairs that fit these criteria without repetition of either protein in the pair in the analysis (Figure 3G, right). This set of pathway partners showed a starkly different pattern of translation- and protein-level correlation than was seen for heterodimers (examples in Figures 3E and 3F, right side of 3G and 3H, and S3A–S3C). In two cases, Trp3/Trp5 and Lys4/Lys12, expression correlation at both translation and protein levels was high and roughly equivalent. Only 7 of the remaining 19 (37%) cases showed a higher protein correlation than translation correlation for the pairs. Subsequent analysis of these cases revealed that at least two of these sets (Erg25/Erg26 and Erg6/Erg2), although not annotated as heterodimers in the Saccharomyces genome database, are in fact reported to associate physically in a complex (Baudry et al., 2001; Mo and Bard, 2005). The remaining 12 gene pairs (57%) showed a poorer correlation between protein levels compared to translation levels (Figures 3G, right, 3H, S3A, and S3B). Only the difference between protein correlations for the two types of gene pairs is significant (p < 0.001), suggesting that the high protein-level correlation observed for complex components is not simply an artifact of our measurements (Figure 3H). We further confirmed that differences in fold changes, absolute changes, or average gene expression levels for gene pairs measured did not skew protein and translation correlation coefficients in our data (Figures S3D–S3G and S4A–S4F).

**Poorly Matched Protein Complex Component Levels Suggest Regulatory Roles**

Of the 39 heterodimers that we examined in the full dataset, only 4 (13.3%) showed protein levels that were more poorly correlated over meiosis than translation levels (Figure 3G). This set of genes is candidates for future study of potentially interesting biological regulation in meiosis, as this is the pattern seen for the Ndc80/Nuf2 heterodimer (Figure 3G), which reflects the recent finding that Ndc80 complex activity is silenced early in meiosis by specific downregulation of Ndc80 protein levels (Chen et al., 2017; Chia et al., 2017; Miller et al., 2012). The downregulation of Ndc80 results in an inactive outer kinetochore during meiotic prophase because Ndc80 acts as a linchpin component. When it is present, the complex is active; when it is not present, the complex is not (Chen et al., 2017; Chia et al., 2017; Miller et al., 2012). When we analyzed their protein levels side by side, as determined by mass spectrometry, all four Ndc80 complex components showed patterns that are consistent with this reported regulation (Figure S5A). Spc24, Spc25, and Nuf2 showed similar protein-level trends to one another over meiosis, but Ndc80 alone showed low levels early in meiosis that rose before the first meiotic division. At the translation level, there was poorer correlation among Spc24, Spc25, and Nuf2 than was seen at the protein level, while Ndc80 was similarly poorly correlated with all three at both the protein and translation levels (Figure S5A).

**Trends in Regulation of Heterodimers Also Apply to Multiprotein Complexes**

Analysis of stable multisubunit protein complexes for which we were able to quantify most or all of the components also revealed evidence of degradation of free monomers (Figures 4A–4E, 4I, 4J, S5B, and S5C). Comparison of translation and protein patterns over meiosis for the oligosaccharyltransferase (OST), chaperonin containing TCP-1 (CCT), F1F0 ATPase, histone deacetylase (HDA), endoplasmic reticulum membrane protein complex (EMC), Ccr4-Not, exosome, translocon, and prefoldin complexes revealed a trend similar to that observed for heterodimers (Figures 4A–4E, 4I, 4J, S5B, and S5C). Although in some cases expression was already well correlated at the translation level, in all of the cases analyzed, the correlation was even...
higher at the protein level. A control comparison of linear biosynthetic pathway members did not show a similar trend, as exemplified by the heme, pyrimidine, ergosterol, histidine, and nucleotide biosynthesis pathways (Figure 4F–4H, S5D, and S5E). Interestingly, although most complex members showed better matching of protein than translation level patterns, outlier components were also observed in two cases analyzed (the exosome and the Ccr4-Not complex), as seen for the Ndc80 complex (Figures 4I, 4J, and S5A). This suggests a mechanism for meiotic regulation of complex activity through regulation of the levels of one or more key members, as discussed above for the Ndc80 complex.

A more detailed investigation of the exosome components revealed that the three protein components (Mpp6, Rrp6, and Lrp1) that did not show high protein level correlation with the others are known to be non-constitutive complex members (Figures 4J and S5A). Mpp6 is an accessory component of the nuclear exosome, and Rrp6/Lrp1 associate tightly as a subcomplex that also associates specifically with the nuclear exosome (Feigenbutz et al., 2013; Synowsky et al., 2009). These three genes clustered tightly with the core exosome components at the mRNA and translation levels, but they are clearly distinct from core exosome components in protein-level clustering, with the closely physically associated Rrp6/Lrp1 showing much tighter protein-level correlation with each other than at either other level of expression measurement (Figure S5A). These patterns support specific post-translational adjustment of protein levels based on physical association of protein complex components.

Analysis of proteasome components similarly revealed a specific protein-level discordance between core and regulatory...
subunits. We were able to quantify all 33 26S proteasome subunits at every level in our gene expression dataset, as well as 4 associated factors. All 26S proteasome subunits showed highly similar patterns of mRNA accumulation and translation, which is consistent with their transcriptional co-regulation (Figures 5B and 5C) (Mannhaupt et al., 1999; Sato et al., 2009). It was not possible, based on clustering of mRNA or translation measurements, to distinguish between subcomplexes (Figures 5B and 5C). In contrast, the protein-level patterns for this set of 33 26S genes showed 2 major groupings, which corresponded almost perfectly with the 19S regulatory particle and the 20S proteasome core (Figures 5B and 5C). The 19S components showed a decrease in protein levels that precedes the decrease seen for 20S subunits by ~3 hr during the meiotic program (Figures 5B and 5C). The core proteasome can be activated independent of the 19S regulatory particle (Schmidt et al., 2005), and the regulation observed late in the meiotic program suggests a natural context for this role. The sole exception to the distinct patterns of protein expression for the core and regulatory subunits was Pre9, which clustered with the regulatory particle subunits rather than the proteasome core, of which it is a reported member (Figure 5B). Pre9 has been reported to be unlike other 20S components in multiple studies. It is known that Pre9 directly interacts with the tails of Rpt2 and Rpt6 (the 2 proteins that it clusters between in Figure 5B) to mediate 19S and 20S association, and the interaction with Rpt6 is the basis for specificity of the 19S/20S registry (Park et al., 2013). Pre9 is also the only non-essential 20S member (Kusmierzczk et al., 2008; Velichutina et al., 2004), suggesting a role that differs from the other core members. It is unclear what this role is, but it is worth noting that cells deleted for PRE9 have been shown to have specific meiotic defects, attributed to the proteasome’s association with meiotic chromosomes during recombination (Ahuja et al., 2017).

All of the proteasome-associated proteins analyzed that did not fit into one of the two major clusters are known to have additional roles (Figure 5B). The 19S regulatory particle component Sem1 is known to be part of other, non-proteasome complexes (Kragelund et al., 2016), and neither Cic1 or Blm10/PA200 are thought to constitutively or exclusively associate with the proteasome. Cic1 associates with the proteasome but also with pre-ribosomal particles (Hampichamchai et al., 2001; Jäger et al., 2001), and Blm10/PA200 is involved in both core particle assembly and mature particle activation (Fehlker et al., 2003). The protein-level distinctions among proteasomal genes are interesting in light of their high degree of known transcriptional co-regulation, and this suggests robust post-translational adjustment of protein levels; but it also demonstrates that protein measurements taken over time can be surprisingly sensitive in identifying functional distinctions among groups of proteins with identical control of synthesis.

Ribosomal Proteins Are Highly Co-regulated at All Levels, Degraded Late in the Meiotic Program

The ribosome is another large multiprotein complex with known co-regulation of component synthesis. In contrast to the proteasome, clustering of ribosomal protein (RP) levels did not reveal subclustering based on any reported physical feature of the ribosome or large versus small subunit identity (Figure 6A). Rather, 72 of the 98 RP-encoding genes that we quantified at the protein level showed extremely similar patterns over the meiotic program at every level of gene expression measured. The other 26 showed slightly different protein-level patterns, the basis of which we do not yet understand (Figure 6A). We also noted that protein-level patterns for all RP genes examined indicated protein degradation late in MI and re-synthesis in spores (Figure 6A). Comparison of the mRNA, translation, and protein abundance measurements for this group of genes revealed a signature of protein degradation similar to what we observed for the known degradation target Sic1 (Figure 1D), with translation actually increased in late meiosis while protein levels decrease (Figure 6B). In spores, subsequently, protein levels increased to a level similar to early meiotic cells, while translation (and mRNA levels) remained high (Figure 6B). While the transcriptional uptick in RP genes had previously been seen during sporulation (Chu et al., 1998), its association with protein degradation was not evident.

To confirm this regulation independently, we GFP-tagged the RP gene RPL26B in a strain carrying mCherry-tagged histone H2B (encoded by HTB1). Both tags were heterozygous in diploid cells. Thus, during meiotic stages in which the cytosol was continuous, before spore packaging, we expect to see homogeneous green cytosolic and red nuclear signals. Following spore individualization, we would expect to see red signal remaining in all four spore nuclei if histones are stable, suggesting that they were synthesized before spore packaging (Figure 6C). In contrast, if RPs were degraded and re-synthesized in spores, then only the two spores carrying the RPL26B-GFP allele should be green and the other two spores should lose GFP signal relative to earlier stages (Figure 6C). Indeed, this was what we observed. We saw evidence that Htb1 continues to be synthesized in spores, resulting in an increase in signal in the two spores that presumably carry the HTB-mCherry allele, but in the case of RPL26B, we observed an increase in signal in two spores and a decrease in the other two (Figures 6C and 6D).

Figure 5. Among Highly Correlated Protein-level Trends for Complex Members, Outliers Suggest Non-constitutive Association

(A) Hierarchical clustering of levels of exosome complex components (from Figure 4J) for mRNA (left), translation (middle), and protein (right). Mpp6 (red) is a non-constitutive component and clusters far from others at the protein level, Rrp6/Lrp1 (green) is a non-constitutive heterodimer; these genes closely associate with each other but separate from the exosome complex at the protein level.

(B) Hierarchical clustering of protein data for proteasome components and accessory factors. Matched mRNA (far left) and translation levels (middle). Note two discrete protein-level clusters—one with all 20S components except Pre9 and the other cluster with all 19S components. Far right, a proteasome illustration is color coded to match gene names to its left.

(C) All 19S regulatory (orange) and 20S core (blue) proteasome members were analyzed together, scaled to the maximum value measured for each. The means and SDs (bars) are shown for mRNA (left), translation (middle), and protein (right). Note protein divergence between the two groups of genes at late time points (represented by gray box), suggesting synthesis co-regulation for all but independent post-translational adjustment for two complexes at late stages. See also Figure 4.
High degradation and low synthesis of Gene1 after spore formation
Active degradation and re-synthesis of Gene1 after spore formation

Cell Reports 25, 3603–3617, December 26, 2018 3613

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This was observed and quantified for individual cases and was a general trend among cells of this genotype (Figures 6D and 6E). The loss of GFP signal in two spores was not due to photobleaching resulting from time-lapse imaging, as cells on the same microfluidic plate that were not previously imaged showed similar relative levels of GFP in a 2:2 bright:dim ratio as those that were imaged over a period of time (Figure S6). A similar effect could be seen in cells carrying a heterozygous RPL29-GFP allele (Figures 6E and S6). We concluded that ribosomes are actively degraded and re-synthesized at the end of the yeast meiotic program.

**DISCUSSION**

Important individual examples of regulated protein degradation during meiotic differentiation are well characterized, but it has been challenging to determine the pervasiveness of this mode of regulation in meiosis. We report signatures in matched global quantitative mass spectrometry and ribosome profiling data that allowed us to identify specific, natural, and previously unidentified cases of regulated protein degradation during the yeast meiotic program. These signatures include periods of stable or even increased translation of a given gene, while protein abundance decreases, as well as periods of particularly rapid decline in protein levels for groups of genes in concert. These trends are sensitively detected in our dataset because cells undergoing meiotic differentiation do not display the type of dilution due to cell growth and division that is a major contributor to protein-level decline during mitotic growth (Christiano et al., 2014).

Ama1, a meiosis-specific APC/C subunit, is important for meiotic progression (Cooper et al., 2000). We observed that all three previously identified Ama1 targets that we were able to quantify in our mass spectrometry dataset showed patterns of protein-level change during meiosis that were strikingly similar to one another as well as to a small group of other proteins (Figure 1E). We hypothesized that this group included new Ama1 targets and confirmed that Pes4 and Mip6, two members of this cluster with meiotic roles during a precise window, are degraded in an Ama1-dependent manner (Figure 2). This is an unorthodox approach for the identification of potential E3 targets, but at least in this case, it seems to allow specific and sensitive detection based on the ability to follow natural protein patterns over time in an unperturbed system.

Analyses of protein data also revealed previously unrecognized, coordinated degradation and re-synthesis of RP subunits following gamete (spore) formation (Figure 6). Why would cells expend energy to degrade RPs and concomitantly re-synthesize them? We propose two explanations. First, this could be a mechanism of cellular quality control. It has been shown that the abnormal nucleolar morphology observed in aged yeast cells is reset in all four gametes by the process of meiotic differentiation (Unal et al., 2011). It is possible that this nucleolar morphology reflects defective rRNA synthesis or processing, and thus resultant ribosomes may be of poor quality. Because gamete quality is important for an organism’s genome stability on an evolutionary scale and the proteins synthesized in a gamete provide critical functions, including mediating gene expression, the destruction and re-synthesis of a gamete’s ribosomes may be a mechanism of ensuring gamete integrity. Second, it is possible that ribosome composition or modification is altered in meiosis relative to mitotic growth and that these alterations must be reset after meiosis. The translation of upstream ORFs (uORFs) within 5′ leaders of thousands of mRNAs is dramatically upregulated during meiosis, even in cases in which an apparently identical transcript is present under mitotic and meiotic conditions (Brar et al., 2012). If meiotic modification to the core translation machinery contributes to this effect, then destruction of this machinery could enable cells to return to mitotic translation patterns.

The most surprising finding from our analyses was that members of stable protein complexes are typically synthesized with imprecise stoichiometry that is adjusted post-translationally (Figures 3, 4, and 5). Effectively perfect translation matching is occasionally seen in our dataset, including for the Tub1/Tub2 heterodimer, demonstrating that cells are capable of very precisely matched synthesis levels, even for proteins encoded from genomically distant genes with dissimilar regulatory regions (Figures 3A and 3G), yet this is not the norm (Figures 3, 4, and 5). This conclusion differs from that of recent studies that reported matched synthesis as the rule among protein complex components (Li et al., 2014; Taggart and Li, 2018). There are significant biological and analytical differences between our study and theirs. The conclusion that synthesis rates were matched for protein partners was determined using estimation of absolute translation levels from ribosome profiling data of cells in rich, steady-state growth conditions (Li et al., 2014; Taggart and Li, 2018). Our study, in contrast, was of cells undergoing meiotic...
differentiation. Natural developmental processes, including meiosis, involve proteome remodeling over time, and this may influence cellular strategies for protein complex regulation. Furthermore, our study used analyses of trends in matched series of protein and translation measurements to enable the quantitative comparison of these two different types of data. Despite these key differences, we argue that the data in all of these studies are actually consistent with our conclusions; we find that some heterodimers do show “perfect” synthesis matching and most show a positive correlation between translation patterns over time. It is not that we observe decoupled synthesis rates for complex components, but simply that protein-level patterns match better than translation patterns for most cases analyzed here, suggesting an important role for both levels of regulation in eukaryotes.

Post-translational adjustment of levels of protein complex partners has been previously observed in a variety of experimental systems by perturbation of normal cellular homeostasis through gene (or chromosome) overexpression (Abovich et al., 1985; Dephoure et al., 2014; Gorenstein and Warner, 1977; Ishikawa et al., 2017; Lam et al., 2007; Papp et al., 2003; Sung et al., 2016a; Torres et al., 2010; Warner, 1977; Warner et al., 1985). Unincorporated complex components have been shown to be unstable in several contexts as a result of proteasome-dependent degradation. For example, if a single RP is overexpressed relative to others, then excess subunits are degraded by a mechanism that depends on the E3 ligase Tom1 (Sung et al., 2016b). It has previously been challenging to determine, however, whether the post-translational adjustment of levels of protein complex components occurs naturally in wild-type, unperturbed cells, or simply occurs as a fail-safe for mutant conditions. Our study argues that the former is the case.

The extremely high degree of protein-level correlation for most protein complex components also enables the sensitive detection of outlier components, which may be involved in regulatory roles. For example, the Ndc80 kinetochore complex components are well matched at the protein level, with the exception of the namesake component Ndc80, which is late in meiosis and increases before the first meiotic division (Figure S5A). The mechanism and importance of this regulation during meiosis have been shown (Chen et al., 2017; Chia et al., 2017; Miller et al., 2012). In such cases, a single outlier component can act as a linchpin for complex activity, an efficient mechanism for rapid complex activation or inactivation. Analysis of similar outlier cases for other complexes seems promising for uncovering other similar types of regulation (Figures 4I, 4J, 5, and S5B) (de Lichtenberg et al., 2005).

The findings reported here regarding protein complex component regulation are informative in considering the cellular balance between perfection and efficiency. Yeast cells undergo meiosis in the absence of glucose or amino acids, and this is therefore a context in which cellular economy of resources is extremely important. The prevalence of imprecise synthesis of protein complex partners and subsequent degradation of unpartnered subunits even in these conditions implies a general advantage to this strategy, relative to perfect synthesis matching. Cases such as the tubulin heterodimer reveal that eukaryotic cells are capable of nearly perfect co-synthesis of partner proteins, and the rarity of this regulation suggests something interesting about the cases in which it does occur. In the case of Tub1/Tub2, for which it is known that stoichiometric imbalance leads to toxicity (Burke et al., 1989), the disadvantage of post-translational buffering seems clear. Investigation of the few other cases that we identified as showing extremely well-matched synthesis may yield new insights into the features and cellular roles of these components, as well as broader principles of cellular resource management in gene expression regulation.

STAR METHODS

Detailed methods are provided in the online version of this paper and include the following:

- Key Resources Table
- Contact for Reagent and Resource Sharing
- Experimental Model and Subject Details
  - Yeast growth and sporulation
- Method Details
  - Western blotting
  - Meiotic cell staging
  - Heterodimer analyses
  - Linear pathway analyses
  - Heterozygous RP-GFP imaging
- Quantification and Statistical Analysis
  - Statistics
  - Heterozygous RP-GFP quantification
- Data and Software Availability

SUPPLEMENTAL INFORMATION

Supplemental Information includes six figures and one table and can be found with this article online at https://doi.org/10.1016/j.celrep.2018.12.008.

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AUTHOR CONTRIBUTIONS

G.A.B. and M.J. conceived most of the aspects of this study. The experiments were performed by A.R.E., A.H., S.H., and A.K. The analyses were performed by A.H., A.R.E., M.J., and G.A.B. The manuscript was written and edited by A.H., A.R.E., M.J., and G.A.B.

DECLARATION OF INTERESTS

The authors declare no competing interests.
REFERENCES


**STAR METHODS**

**KEY RESOURCES TABLE**

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**CONTACT FOR REAGENT AND RESOURCE SHARING**

Further information and requests for resources and reagents should be directed to and will be fulfilled by the Lead Contact, Gloria Brar (gabrar@berkeley.edu).

**EXPERIMENTAL MODEL AND SUBJECT DETAILS**

**Yeast growth and sporulation**

All yeast strains used were diploid *Saccharomyces cerevisiae* of the SK1 background. Strains used in this study are listed in the key resources table. For meiotic time courses, yeast were inoculated into YEPD overnight, then diluted to OD600 0.2 into buffered YTA and grown for 12 hours. Cells were washed in water and resuspended in sporulation media supplemented with 0.02% raffinose. Time points were taken at indicated times.

**METHOD DETAILS**

**Western blotting**

Western blotting was performed using a standard TCA-based protocol. Briefly, 2.5 OD units of culture were treated with 5% TCA at 4C for at least 10 min. Samples were then washed with 1 mL acetone. Acetone was aspirated and pellets were dried overnight at RT. Lysates were made by adding 100 µL protein lysis buffer [50 mM TE, 3 mM DTT, 1.1 mM PMSF (Sigma), 1 µM pepstatin A,
1X protease inhibitor cocktail (Roche) and 1 volume acid-washed glass beads (Sigma), and bead-beating for 5 min at RT. 3X SDS loading buffer was added and samples were boiled for 5 min. Beads were pelleted by centrifugation and 5 µL supernatant was loaded onto 4–12% Bis-Tris polyacrylamide gels. Following electrophoresis, proteins were transferred using a semi-dry transfer apparatus (Trans-Blot Turbo, BioRad). The following antibodies were used: mouse anti-V5 (Invitrogen, 1:2,000), rabbit anti-hexokinase (Stratech, 1:10,000), anti-mouse and anti-rabbit secondaries (Li-Cor, 1:15,000). Primary antibody incubation was overnight, secondary for 1-2 hr. Blots were visualized and quantified using a Li-Cor system.

**Meiotic cell staging**
The meiotic stage of a cell was determined based on its DAPI morphology by fluorescent microscopy. 200 cells were counted per strain per time point.

**Heterodimer analyses**
Heterodimers were defined as annotated in the *Saccharomyces* Genome Database (SGD; https://www.yeastgenome.org). Only cases in which both partners were quantified for mRNA, translation, and protein were analyzed and cases were excluded in which there was a reported major alternative interactor (typically a paralog). Three additional cases were excluded for which one partner appeared to be especially lowly expressed (a mean translation RPKM < 20). Z-scores were calculated and the correlation over all 10 samples for the pair of genes was determined.

**Linear pathway analyses**
Linear biosynthetic pathways were identified based on literature searching and SGD confirmation. Only cases in which both partners were quantified for mRNA, translation, and protein were analyzed and cases were excluded in which there was a reported major alternative interactor (typically a paralog). Also, each gene was only analyzed in one pairing (either with its upstream or downstream partner, decisions about which to use were based on maximization of possible partners to analyze). Unlike the heterodimer analyses, no cases needed to be excluded based on especially low expression of either gene (a mean translation RPKM < 20). Z-scores were calculated and the correlation over all 10 samples for the pair of genes was determined.

**Heterozygous RP-GFP imaging**
After two hours in SPO media, 100 µL of cells were placed in a CellASIC ONIX Microfluidic Plate (Y04D) and maintained with fresh SPO media at 2psi using the CellASIC ONIX Microfluidic Perfusion System (CellASIC Corp., Hayward, CA, USA). The cells were held at 30 C using a thermostatic system for the microscope stage.

Cells were imaged using a DeltaVision microscope with a 60x/1.42 oil-immersion objective (DeltaVision, GE Healthcare, Sunnyvale, CA) and filters: FITC (EX475/28, EM525/48) and mCherry (EX575/25, EM625/45). Images were acquired using the softWoRx software (softWoRx, GE Healthcare) with z stacks of 8 slices with 0.5 µm spacing. For time lapse imaging, images were taken every 20 minutes for 12 hours. After 24 hours, images were taken from the same points that had been imaged during the time-lapse portion, as well as points from the same wells that had not been imaged previously.

**QUANTIFICATION AND STATISTICAL ANALYSIS**

**Statistics**
A Kolomogorov-Smirnov (K-S) test was used to determine significance for the differences between cumulative distribution plots shown in Figure 3H. A two-tailed t test was used to determine significance between fluorescence intensities for the data presented in Figure 6E.

**Heterozygous RP-GFP quantification**
All images were deconvolved using softWoRx software accompanying the DeltaVision microscope, and maximum-intensity projections were generated using ImageJ/FIJI image processing software (RRID:SCR_002285; Schneider et al., 2012). Mean intensity of signal from the cells was measured using the “measure” tool in FIJI, and was divided by the background signal from the same image.

**DATA AND SOFTWARE AVAILABILITY**

Genome-wide data analyzed here were generated previously (Cheng et al., 2018). In short, mRNA levels were assayed by mRNA-seq, translation measurements were assayed by ribosome profiling, and protein levels were assayed by quantitative mass spectrometry (TMT10) for 8 time points during the meiotic program and two vegetative time points (one in rich media and one in sporulation media matched to meiotic samples). All measurements showed high reproducibility, with R values ranging from 0.935 to 0.992. All original data can be found at NCBI GEO: GSE108778 and MassIVE: MSV000081874. Processed data used for analyses here are in Table S1.
Supplemental Information

Precise Post-translational Tuning Occurs for Most Protein Complex Components during Meiosis

Amy Rose Eisenberg, Andrea Higdon, Abdurrahman Keskin, Stefanie Hodapp, Marko Jovanovic, and Gloria Ann Brar
Supplemental Figures:

Figure S1: Meiotic staging and replicate for experiment pictured in Figure 2, Related to Figure 2. A-E) Cells were staged by DAPI quantification at hourly timepoints (n=200 per strain per timepoint). Note a slight delay in cells deleted for AMA1 in all cases that can not account for the larger delay in protein depletion in these strains. A) Progression data for experiment shown in 2D B) Progression data for experiment shown in 2E C) Progression data for experiment shown in 2F D) Progression data for experiment shown in 2G E) Progression data for experiment shown in 2H. F-J) Replicate time courses were run for experiments shown in Figure 2D-2I and western blotting and quantification were performed. F) Replicate for experiment shown in 2D G) Replicate for experiment shown in 2E H) Replicate for experiment shown in 2F I) Replicate for experiment shown in 2G J) Replicate for experiment shown in 2H K-O) Cells were staged as in S1A-E. Note a slight delay in
cells deleted for *AMA1* in all cases that can not account for the larger delay in protein depletion in these strains. K) Progression data for experiment shown in S1F L) Progression data for experiment shown in S1G M) Progression data for experiment shown in S1H N) Progression data for experiment shown in S1I O) Progression data for experiment shown in S1J
Figure S2: Protein-level clustering of genes translated during recombination and SC formation reveals interaction- and function-related substructure, Related to Figures 3 and 4. At left is the translation pattern for a group of 46 genes.
that co-clustered translationally in our original meiotic ribosome profiling study (Brar et al., 2012). At right, clustering of the protein-level data for the 35 of these genes that were quantified by mass spectrometry. Note that bold names represent genes with known relationships to recombination and SC formation. Also note that groups of genes with similar function or similar physical interactions tend to cluster together at the protein level but not the translation level. This effect is especially striking for the four genes that encode so-called “Zmm proteins”, with functions in linking recombination and SC formation (Lynn et al., 2007), which cluster tightly together at the protein level and consist of two heterodimers, which cluster even more tightly than the group as a whole.
Figure S3: Sequential enzyme pairs and heterodimer partners behave differently, but correlation differences are not associated with differences in fold-changes, Related to Figure 3. A-C) Alternative representations of the data in
Fig. 3G, 3H. A) The correlation between heterodimer pairs (purple) and adjacent enzymes (turquoise) is plotted with the x-axis representing translation and the y-axis representing protein. Note that the former skew heavily towards the upper right (and particularly very high y values), while the latter skew rightward, representing high correlation of synthesis only. B) A cumulative distribution plot of the data in Fig. 3G is shown comparing translation and protein correlations among adjacent enzymes in biosynthetic pathways. The difference is not significant by K-S test. C) A cumulative distribution plot of the data in Fig. 3G is shown comparing translation and protein correlations among heterodimer pairs. The difference is significant by K-S test (p-value= 0.001). D-G) Differences in the degree of regulation are not associated with skewed correlation values. D) Fold changes from highest to lowest values measured for translation for the set of adjacent pathway genes shown in Fig. 3G are plotted against translation correlation coefficients and no significant association is seen. E) The same analysis for the heterodimers in Fig. 3G is shown. Note a propensity for very highly regulated pairs to show a high correlation in translation, likely due to highly similar mechanisms of regulation of synthesis and evolution for cellular efficiency. The outlier here is, for example, Msh4/Msh5, which are expressed in a highly meiosis-specific manner. No similarly highly regulated (or meiosis-specific) biosynthetic pathway genes were found to allow comparison. F) Fold changes from highest to lowest values measured for protein for the set of adjacent pathway genes shown in Fig. 3G are plotted against protein correlation coefficients and no significant association is seen. G) The same analysis for the heterodimers in Fig. 3G is shown. Note a propensity for very highly regulated pairs to show a high correlation in protein abundance, likely due to highly similar mechanisms of regulation of and evolution for cellular efficiency. No similarly highly regulated biosynthetic pathway genes were found for comparison (see note above for E).
Figure S4: Levels of expression do not skew correlation analyses for the genes analyzed in our study, Related to Figure 3. A) Absolute average levels of translation and protein for the set of adjacent pathway genes shown in Fig. 3G are plotted against the appropriate translation or protein correlation coefficients and no
significant association is seen. B) Absolute average levels of translation and protein for the set of heterodimer pairs shown in Fig. 3G are plotted against the appropriate translation or protein correlation coefficients and no significant association is seen. C) Absolute changes in translation level measured in our dataset for the set of adjacent pathway genes shown in Fig. 3G are plotted against translation correlation coefficient and no significant association is seen. D) The same analysis is shown for the set of heterodimer pairs shown in Fig. 3G. E) Absolute changes in protein level measured in our dataset for the set of adjacent pathway genes shown in Fig. 3G are plotted against protein correlation coefficient and no significant association is seen. F) The same analysis is shown for the set of heterodimer pairs shown in Fig. 3G. Note that few poorly correlated protein pairs are available for analysis and none show evidence for big shifts in absolute levels, but that small absolute changes alone do not preclude or associate with any particular correlation coefficient.
Figure S5: Additional examples of differences in translation- and protein-level trends among complex and biosynthetic pathway members, Related to Figures 4-6. For all panels A-E, translation levels (z-score) are plotted at the left and protein at the right. A) Plotted are data for the Ndc80 complex members. The regulation of this subcomplex is well characterized and it is established that Ndt80 synthesis alone is delayed in meiosis to limit kinetochore activity (Chen et al., 2017; Chia et al.,...
2017; Miller et al., 2012). Our data recapitulate this result. Translation data do not match well for any of the four subunits, but protein data trends match well for Nuf2, Spc24, and Spc25, while Ndc80 accumulation is delayed relative to the others. B) Plotted are data for the translocon complex components. Note that Sec72 is the only outlier at the protein level and is the only non-essential component of this complex. C) Plotted are data for the prefoldin complex components. D) Plotted are data for the histidine biosynthesis pathway components. E) Plotted are data for the purine nucleotide biosynthesis pathway.
Figure S6: The low signal in two spores without fluorescently-tagged RP genes is not due to photo-bleaching, Related to Figure 6. Average integrated intensity of the two brightest or two dimmest spores from tetrads are compared for strains carrying heterozygously tagged Rpl26b and Htb1 (left) or Rpl29 and Htb1 (right). Plotted values are either for cells imaged after 50 frames or for not previously imaged cells at the same time point. 10 tetrads were counted per strain.