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**The anti-cancer drug 5-fluorouracil affects cell cycle regulators and potential regulatory long non-coding RNAs in yeast**

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**Short title:** 5-fluorouracil affects transcriptional regulators and lncRNAs in yeast

## Abstract

5-fluorouracil (5-FU) was isolated as an inhibitor of thymidylate synthase, which is important for DNA synthesis. The drug was later found to also affect the conserved 3'-5' exoribonuclease EXOSC10/Rrp6, a catalytic subunit of the RNA exosome that degrades and processes protein-coding and non-coding transcripts. Work on 5-FU's cytotoxicity has been focused on mRNAs and non-coding transcripts such as rRNAs, tRNAs and snoRNAs. However, the effect of 5-FU on long non-coding RNAs (lncRNAs), which include regulatory transcripts important for cell growth and differentiation, is poorly understood. RNA profiling of synchronized 5-FU treated yeast cells and protein assays reveal that the drug specifically inhibits a set of cell cycle regulated genes involved in mitotic division, by decreasing levels of the paralogous Swi5 and Ace2 transcriptional activators. We also observe widespread accumulation of different lncRNA types in treated cells, which are typically present at high levels in a strain lacking EXOSC10/Rrp6. 5-FU responsive lncRNAs include potential regulatory antisense transcripts that form double-stranded RNAs (dsRNAs) with overlapping sense mRNAs. Some of these transcripts encode proteins important for cell growth and division, such as the transcription factor Ace2, and the RNA exosome subunit EXOSC6/Mtr3. In addition to revealing a transcriptional effect of 5-FU action via DNA binding regulators involved in cell cycle progression, our results have implications for the function of putative regulatory lncRNAs in 5-FU mediated cytotoxicity. The data raise the intriguing possibility that the drug deregulates lncRNAs/dsRNAs involved in controlling eukaryotic cell division, thereby highlighting a new class of promising therapeutical targets.

**Key words:** 5-fluorouracil, transcription factor, *SWI5*, *ACE2*, *AMNI*, Mitotic Exit Network, long non-coding RNAs, antisense RNA, double stranded RNA, chemotherapy, yeast

## Introduction

In spite of substantial progress made over the past couple of decades, effective chemotherapeutical treatment of most somatic cancers remains a challenge. Within a large variety of anti-cancer drugs available, agents that inhibit DNA replication and repair are among the oldest and most widely used compounds in therapies of many solid tumors. One of the best-studies examples, 5-fluorouracil (5-FU), was initially designed in 1957 to interfere with pyrimidine biosynthesis. Later work confirmed its ability to inhibit a key enzyme in this pathway, thymidylate synthase, and to interfere with the formation of ribosomal RNA [1, 2, 3]. Subsequently, 5-FU was proposed to also act through an RNA pathway [4, 5]. The mechanism of action involves incorporation of fluorinated uracil into transcripts, **which converts them into poor substrates for** the conserved 3'-5' exoribonuclease EXOSC10/Rrp6 [6, 7, 8]. EXOSC10/Rrp6 is involved in global RNA processing and degradation either independently of or in association with the RNA exosome, which is a multi-subunit complex conserved from yeast to human [9, 10].

*Saccharomyces cerevisiae* is a versatile model organism that has been used extensively to study mitotic cell growth and division in the presence of anti-proliferative compounds such as 5-FU [11, 12, 13]. While both DNA- and RNA based mechanisms have been invoked for the drug's effects in cancer cells, it is noteworthy that 5-FU toxicity in *S. cerevisiae* appears to be mostly due to altered RNA processing [5, 13]. This marks yeast out as a useful system to study the global transcriptional response to 5-FU. Yeast *RRP6*, although not annotated as an essential gene, is important for normal cell growth and differentiation in various strain backgrounds [14, 15, 16, 17]. Its human ortholog *EXOSC10* was initially thought to be non-essential [18], but more recent work demonstrated its critical role for somatic and germ cell division in mouse and human [19, 20, 21].

A large fraction of EXOSC10/Rrp6's substrates are lncRNAs that are broadly defined as transcripts with little or no coding potential. In yeast, this class of RNAs includes cryptic unstable transcripts (CUTs) and meiotic unannotated transcripts (MUTs) that accumulate to peak levels in a vegetatively growing *rrp6* mutant and sporulating wild type cells [16, 22]. Two other major types of lncRNAs are Xrn1-sensitive unstable transcripts (XUTs), and stable unannotated transcripts (SUTs) [23, 24]. The roles of most lncRNAs are unknown. However, a growing body of evidence indicates that many of them are not a dispensable byproduct of bi-directional transcription, but regulate gene expression *via* promoter

interference, and sense/antisense (s/a) interference of transcription in both fission- and budding yeast [25, 26, 27, 28, 29, 30, 31].

Regulatory lncRNAs are known to play important roles also in higher eukaryotes, notably during mammalian cell growth and differentiation [32]. Specifically, it was shown that a dsRNA molecule formed by s/a lncRNAs has important functions in the human cell cycle [33]. Such interactions between transcripts are likely not an isolated phenomenon, since recent genome-wide profiling methods revealed an unexpectedly complex RNA interactome in mammals, which includes numerous antisense lncRNAs [34, 35, 36]. lncRNAs are not only interesting from a biological standpoint, they currently also receive much attention as drivers of cancer progression [37, 38]. Interestingly, 5-FU induces genes in the interferon pathway, and dsRNAs formed by retrotransposons mediate an interferon response in cancer cells, when DNA methylation is inhibited [39, 40]. It is, however, unknown if dsRNAs that originate from s/a transcript pairs are also able to trigger an immune response to cancer.

Given that EXOSC10/Rrp6 targets lncRNAs in dividing cells, it is conceivable that exoribonuclease-dependent 5-FU toxicity is in part mediated by its ability to alter the stability of potential regulatory lncRNAs. To test this idea, we have carried out a genome-wide RNA profiling analysis of 5-FU treated versus untreated diploid wild type cells and *rrp6* mutant control cells synchronously undergoing a mitotic cell cycle. Our study has revealed protein-coding and non-coding transcripts that show altered levels in 5-FU treated cells and *rrp6* mutant cells. Somewhat unexpectedly, we identified the paralogous transcriptional regulators Swi5 and Ace2 as 5-FU targets. These activators are important for the mitotic exit network, a group of genes that orchestrates cell cycle progression and cell separation. The analysis also identified 5-FU responsive non-coding RNAs, including antisense transcripts that form *in vivo* dsRNAs with sense mRNAs involved cell cycle control. Our results therefore reveal an indirect transcriptional effect of 5-FU, and they highlight lncRNAs in general, and antisense transcripts that form dsRNAs with overlapping mRNAs in particular, as potential mediators of the drug's RNA-dependent effects on cell division.

## Results

### *Experimental rationale and design*

The major aim of our study was to analyse the effect of 5-FU treatment as compared to *RRP6* deletion on mRNA and lncRNA levels in diploid yeast cells undergoing rapid mitotic growth. Since 5-FU down-regulates Rrp6's activity via substrate inhibition, we expected the drug to increase the enzyme's target transcript levels. Although RNA profiling studies of the mitotic cell cycle are typically carried out in haploid cells, we preferred using a diploid strain to ensure that the data are comparable to our earlier work on lncRNA and protein expression during cell growth and differentiation [16, 41]. Consequently, haploid cell synchronization protocols based on pheromone arrest were not suitable. We also sought to avoid temperature sensitive mutant alleles of essential cell division cycle genes, since they may have unintended effects on 5-FU's activity, and *rrp6* mutant cells (referred to as *rrp6* $\Delta$ ) fail to grow at 37°C. We therefore arrested wild type cells in G1 by starvation [42], which is a naturally occurring process, before releasing them into fresh rich medium (YPD) in the presence and absence of 20 $\mu$ g/ml 5-FU (Fig 1). As a control for lncRNA accumulation, *rrp6* $\Delta$  cells were arrested and released into rich medium. The strains were sampled at 40 and 100 minutes, which are two time points broadly corresponding to G1/S and the onset of G2/M-phases as determined by monitoring DNA replication. We note that synchronizing cells enhances fluctuating RNA levels, which might be masked in samples from asynchronous logarithmically growing cells. We cannot, however, study *bona fide* cell cycle regulation of coding and non-coding transcripts in the presence and absence of 5-FU, because the experiment does not cover at least two complete mitotic cycles after nutrient deprivation.

### *5-FU treatment of diploid yeast cells*

To optimize 5-FU treatment in our diploid JHY222 strain we assayed growth at different concentrations and found that 20 $\mu$ g/ml closely mimics the growth phenotype of *rrp6* $\Delta$  cells at 30°C. We also confirmed previous findings that cells lacking *RRP6* are hyper-sensitive to 5-FU and that growth at 37°C exacerbates the drug's toxic effect (Fig 2A) [5, 11, 12, 43].

We subsequently monitored progression through G1 and S-phase by FACS analysis of cells sampled every 20 minutes until three hours after release into fresh medium. The data indicated that most cells had not initiated a new round of mitosis 40 minutes (representing

G1/S) into the experiment, and that approximately 50% of the cells had undergone DNA replication by 100 minutes (representing G2/M). 5-FU treated wild type cells and the *rrp6Δ* control show a moderate delay at the 100-minute time point and, as expected, a substantial delay at 180 minutes (Fig 2B) [11].

### *The RNA expression landscape of 5-FU treated diploid cells*

We first compared the global RNA signal levels between the samples using a scatter plot matrix. This analysis revealed, as expected, a strong effect of *RRP6*'s deletion, and a similar albeit much weaker effect of 5-FU on global RNA levels (S1A Fig). A more specific analysis of the transcriptional response shown in density graphs and box plots revealed that CUT levels showed the strongest increase in the absence of Rrp6, and in wild type cells treated with 5-FU, while mRNA expression levels showed the smallest increase. For SUTs, XUTs and MUTs we observed an intermediate response (S1B Fig). These results are consistent with CUTs being a major target of Rrp6, and 5-FU inhibiting the enzyme via substrate inhibition [6, 22].

We next identified genes for which we observed fluctuating levels across the sample set. Given the experimental setup, we expected a transcriptional response related to cells progressing from G1/S toward G2/M together with the effects due to lacking Rrp6 activity and 5-FU treatment. We first identified 9786 RNAs (including 5596 mRNAs, 400 dubious open reading frames (ORFs) and 3651 lncRNAs) that showed an expression value given in counts per million (CPM)  $\geq 1$  in at least four out of 12 samples. Subsequently, we employed TMM normalization and a statistical test implemented in EdgeR corrected for multiple testing (adjusted p-value  $< 0.05$ ) to filter 5697 significantly differentially expressed RNAs (see methods). They include 2617 validated mRNAs, 248 transcripts encoded by dubious ORFs, and 2778 RNAs annotated as non-coding RNAs that fall into five classes (CUTs, SUTs, rsSUTs, MUTs and XUTs). This multi-step procedure was used to decrease the false-positive rate due to very low expression signals (see methods). Finally, we empirically grouped the mRNAs into five clusters using the supervised k-means algorithm. Clusters C1 and C2 contain genes that respond to 5-FU and/or *rrp6Δ*. C3 contains genes for which we observe decreased levels in *rrp6Δ* and 5-FU. C4 and C5 contain mRNAs that tend to peak at the G1/S or G2/M time points, respectively, in all three conditions (Fig 3B). The latter two

patterns are exemplified by the G1 cyclin *CLN2* (C4) and the B-type cyclin *CLB2* (C5) (Fig 3C; S2 Table).

To further analyse the functional relevance of our clustered target transcripts, we identified significantly enriched Gene Ontology (GO) terms from the *biological process* ontology (S2 Figure, S3 Table) and we manually searched the clusters for genes related to cell growth, division and stress response [44]. C1 contained genes important for the mitotic cell cycle and cell division such as a cyclin (*PCL9*) but showed no statistically significant enrichment of these biological processes. We found that C2, for which we also detected no GO term enrichment, contained genes involved in processes relevant for 5-FU and the known phenotype of *rrp6Δ*. They include DNA repair (*HIM1*, *POL2*, *SSL2*), DNA replication (*ORC5*, *PSF1*, *SLD5*), regulation of gene expression relevant for stress response and ribosomal RNA synthesis (*ABF1*, *HCMI*, *MOT3*, *RGT1*, *RRN3*, *RRN7*), and cell cycle progression (*BUB2*, *CDC34*). We also observed genes important for meiosis (*MEI4*, *RED1*) and spore formation (*SPO19*, *SPO73*, *SSP2*) in C2. This is unsurprising, since our experimental protocol involved arresting diploid cells in G1 via starvation, which primes them for entry into the meiotic differentiation pathway (the full execution of which requires, however, the simultaneous absence of a fermentable carbon source and nitrogen; see methods) [45].

Cluster C3 contains genes that fail to be upregulated to normal levels in *rrp6Δ* and in 5-FU treated cells that progress through S-phase toward G2/M. Most of these genes are involved in metabolic processes, notably metabolism (*ENO1*, *GPM1*, *TDH2*) and transport (*OPT1*, *SIT1*). C3 also contains genes important for the cell cycle (*AMN1*, *CTSI*, *DSE1*, *DSE2*, *DSE3*, *DSE4*, *EGT2*). The reasons for this *rrp6Δ*-dominant effect are unclear but they are likely related to the complex nature of the mutant phenotype [46, 47] (S2A Fig).

Cluster C4 identifies genes that show a broadly similar pattern of decreased expression in the 100-minute samples in all three conditions. Enriched processes include DNA-dependent DNA replication (*CDC9*, *CDC45*, *RAD27*, *RFA1*, *RFA2*, *RFA3*, *SIMI*) and DNA damage stimulus (*MSH6*, *RTT107*). We note, however, that some DNA repair genes are induced by 5-FU. Such cases include enzymes important for nucleotide excision repair (*RAD4*), double strand break repair (*PRI2*), recombination (*SMC6*), mismatch repair (*PMS1*), and DNA-damage induced transcription (*DUNI*). These patterns reflect 5-FU's impact on cell cycle progression and DNA integrity [48, 49] (S2B Fig).

In Cluster C5, where we observe consistent up-regulation in the entire sample set, a range of genes involved in amino acid biosynthesis pathways (*ARO3*, *HOM2*, *ILV6*, *MET6*, *MET17*, *MMP1*) appears to be unaffected by *RRP6* gene deletion and drug treatment. It is noteworthy that this cluster also contains cell cycle regulators that are not perturbed at the level of mRNA expression (*CLB1*, *CLB2*, *SPO12*). Although this group is not significantly enriched, it appears that neither 5-FU nor *rrp6Δ* prevent cell cycle regulated genes in general from being expressed in our experimental setup (S2C Fig).

*5-FU affects the mitotic exit network via the activator Swi5*

Cluster C3 contained the Antagonist of the Mitotic Exit Network (*AMN1*), which is involved in mitotic checkpoints and chromosome stability. The gene is down-regulated by 5-FU and (albeit to a lesser extent) by *rrp6Δ* at the 100-minute time point [50, 51] (Fig 4A). We confirmed the RNA-Sequencing data in an independent experiment using RT-PCR (Fig 4B). Furthermore, we observed similar patterns for a cyclin-dependent kinase inhibitor (*SIC1*) and genes involved in daughter cell separation (*CTS1*, *DSE1*, *DSE3*, *DSE4*, *EGT2*; Fig 4C). The distinctive expression pattern of these genes likely explains why they were clustered apart from most other cell cycle regulated genes in C4 and C5 that appeared to be unperturbed at the transcriptional level. These findings prompted us to search for more genes important for cell separation and cell wall synthesis that showed negative 5-FU/*rrp6Δ* response patterns. We indeed found *AIM44* and *PIR1* in C5 (Fig 4C). We note that the transcripts shown in Fig. 4C accumulate to higher levels in *rrp6Δ* cells than in the wild-type at the 40 minute time point. It could be that they are under negative control of Rrp6 during G1/S phase.

The observed patterns could be explained by mechanisms either involving RNA synthesis (gene activation) or RNA degradation (transcript stability). To distinguish between these possibilities we interpreted information about published *in vivo* protein/DNA binding data from microarray-based chromatin immunoprecipitation (ChIP-Chip) experiments that are online available via the YEASTRACT ([www.yeasttract.com](http://www.yeasttract.com)) database [52]. This confirmed that, among others, *AMN1*, *AIM44*, *DSE1*, *DSE2*, *DSE4*, *EGT2*, *PCL9*, *PIR1* and *SIC1* are likely direct targets of the DNA binding activator Swi5 [53, 54, 55]. We therefore examined RNA levels for *SWI5* and found that it was expressed at lower levels in 5-FU than untreated cells at 100 minutes, while no negative effect was apparent in the *rrp6Δ* sample (Fig 5A). This pattern was confirmed at the protein level. Swi5 protein is undetectable under our

experimental conditions in the 40-minute samples, while it accumulates to roughly equivalent levels in WT and *rrp6Δ* cells but not in 5-FU treated WT cells at 100 minutes (Fig 5B). It is unclear why we cannot detect Swi5 at the early 40-minute time point but it might be degraded in diploid cells undergoing glucose starvation and G1 arrest, perhaps by the ubiquitin ligase SCF(Cdc4) that is known to target Swi5 during the cell cycle [56]. Swi5's decrease at the 100-minute timepoint could in part explain the strong negative effect 5-FU has on the expression of its target genes. An additional contribution could be made by partial cytoplasmic retention of Swi5 in treated cells, bearing in mind that the protein is in part regulated via subcellular localisation [57].

#### *5-FU treatment affects different classes of lncRNAs*

Given that 5-FU-treatment leads to the production of modified transcripts, which are inefficient substrates for Rrp6's ribonucleolytic activity we hypothesized that the drug might stabilize potential regulatory lncRNAs [6, 7]. At least some of these RNAs, notably antisense transcripts that form dsRNAs with their overlapping sense mRNAs or with other lncRNAs, possess regulatory functions or have been associated with altered protein levels [41, 58].

To gain insight into the RNA profile of diploid cells treated with 5-FU versus *rrp6Δ*, we first grouped 2778 lncRNAs, including CUTs, SUTs, MUTs and XUTs, into five clusters broadly representing a positive 5-FU/*rrp6Δ* response (C1-C2), a negative 5-FU/*rrp6Δ* response (C3), peak expression in G1/S (C4) and peak expression in G2/M (C5) (Fig 6). These patterns were reminiscent of what we obtained with mRNAs. The overall response is comparable in 5-FU and *rrp6Δ* samples but the lncRNAs accumulate to much higher levels in the gene deletion mutant. This was to be expected since the deletion strain entirely lacks Rrp6 activity, while the enzyme might only be partially affected in the 5-FU treated wild type strain.

Given the regulatory potential of these non-coding RNAs, we next focused on the sub-class of 804 5-FU/*rrp6Δ* responsive antisense lncRNAs that overlapped sense mRNAs (Figure 7A and S4 Table; note that the table contains 825 entries because of multiple overlap cases). GO term analysis revealed no significant enrichment, but we observed genes involved in meiosis (*MEI4*, *PCH2*, *RED1*, *ZIP2*), sporulation (*ADY4*, *SPO22*, *SPO77*) and, notably, cell growth (cell wall organization; e.g. *CWPI*, *EXG2*) (S4 Table). We next asked if these s/a transcript pairs form dsRNAs *in vivo*. To this end, we interpreted data from a published dsRNA profiling experiment using a haploid wild type strain and a mutant strain lacking the 5'-3'-

exoribonuclease Xrn1 (meant to amplify the signal by stabilizing antisense transcripts) that ectopically express fungal Dicer and Argonaute orthologs from *N. castellii*. These enzymes cleave dsRNAs and the breakdown products are detected by small RNA-Sequencing [58]. Using these dsRNA profiling data, we observed clear dsRNA signals for several meiotic genes and, importantly, the cell growth factors *CWPI* and *EXG2* (S3 Figure).

The group of s/a loci also contained a class of genes involved in response to chemicals, such as *MNLI/SUT597* that is relevant for 5-FU treatment. *MNLI* is potentially interesting because it is conserved from yeast to mammals (Mnl1/EDEM1) and plays a role in resistance to chemicals via the unfolded protein response pathway. Its antisense RNA *SUT597* is upregulated in the presence of 5-FU and the absence of Rrp6 as shown both by RNA-Sequencing and an RT-PCR assay (Figure 7B-D). As expected, a stable dsRNA is formed by *MNLI/SUT597* s/a transcripts (Figure 7E). *CUT158/MTR3* is also interesting because this conserved gene (*EXOSC6* in human) is involved in RNA exosome-dependent transcript processing and degradation. *CUT158* accumulates in 5-FU treated and *rrp6* mutant cells as monitored by RNA-Sequencing and RT-PCR assay and forms a dsRNA that covers almost the entire *MTR3* coding sequence (S4 Fig A-D).

Intriguingly, the group of s/a loci includes the *SWI5* paralog *ACE2*, and its overlapping antisense lncRNAs *CUT259* and *SUT2107* that show opposing peak expression patterns both in RNA-Sequencing and stranded RT-PCR experiments (Fig 8A-C). We observed that *ACE2* mRNA is moderately down-regulated by 5-FU but not *rrp6* $\Delta$  (Fig 8A-B). When we measured Ace2 levels by Western blot, we found that the protein failed to accumulate to wild type levels at the 100-minute time point in both treated and mutant cells; this effect became weaker at a later time point at 160 minutes (Fig 8D). This finding raised the question why Ace2 did not accumulate to normal levels in 5-FU treated and *rrp6* $\Delta$  cells although mRNA levels remain almost the same. It is possible that the mechanism, which down-regulates the protein in both conditions involves *ACE2/SUT2107* dsRNA formation, which occurs in haploid cells undergoing growth in rich medium (Fig 8E) [58]. We therefore assayed *ACE2/SUT2107* transcript pairs in our sample set from diploid cells using an immunoprecipitation assay based on the dsRNA-specific J2 antibody and RT-PCR [59]. We detected weak signals in untreated and treated cells, and a stronger signal in the *rrp6* $\Delta$  background that was consistent with the increase of sense and antisense RNA levels in this strain (Fig 8F). Finally, we confirmed that the J2 assay is suitable for detecting fluctuating dsRNA signals due to variable transcript levels by monitoring *GAPI/SUT1912* that are very

weakly expressed in fermenting mitotic cells (YPD) and induced in meiotic cells (cultured in sporulation medium SPII; S5 Fig A-C; the complete dataset will be published elsewhere).

We conclude that antisense SUT2107 levels are increased moderately by 5-FU and strongly by *rrp6* $\Delta$ , thereby drawing *ACE2* sense mRNA into a dsRNA complex. The *ACE2*/SUT2107 double-stranded form covers almost the entire coding region of *ACE2*, and is associated with diminished Ace2 protein levels in 5-FU and *rrp6* $\Delta$  samples. This result complements our data on Swi5, because Ace2 is known to co-activate certain Swi5-dependent cell cycle genes, and other genes important for cell division, such as *CTS1* ([60] and references therein). Our findings indicate that 5-FU negatively affects Swi5 and Ace2 transcription factors that are important for normal cell cycle progression via the Antagonist of the Mitotic Exit Network (MEN) pathway, and its downstream target genes. The data highlight a transcriptional regulatory aspect of 5-FU's toxicity that implicates important cell cycle regulators, and stable lncRNAs/dsRNAs. Finally, the results provide molecular insight into the complex mitotic growth phenotype shown by *rrp6* $\Delta$  cells in different strain backgrounds at permissive and restrictive temperatures [16, 17, 61].

## Discussion

The anti-cancer drug 5-FU is incorporated into RNA and thereby indirectly inhibits the 3'-5' exoribonuclease EXOSC10/Rrp6, which is an evolutionarily conserved protein that processes or degrades different types of non-coding RNAs. Since certain long non-coding RNAs have been reported to be important for cell growth, development and diseases such as cancer, we hypothesized that part of 5-FU's cytotoxicity might stem from its ability to alter the stability of known or putative regulatory lncRNAs. To provide initial evidence for the concept in a model system suitable for RNA profiling of cell cycle progression, we analysed the effect of 5-FU on the transcriptome of synchronized diploid budding yeast wild type versus *rrp6* mutant cells. The latter served as a positive control since cells lacking Rrp6 accumulate lncRNAs, and fail to divide normally; that is, while the phenotype is more or less severe at the permissive temperature depending on the background, all mutant strains arrest at the restrictive temperature of 37°C.

### *EXOSC10/Rrp6 is important for cell growth and development*

We find that the vast majority of 5-FU/*rrp6*Δ dependent lncRNAs accumulate to much higher levels in the gene deletion mutant than in drug-treated cells, yet JHY222 *rrp6*Δ cells can grow and divide. While this seems to argue against a major effect of deregulated lncRNAs on cell division, we underline that JHY222 *rrp6*Δ cells do show a moderate growth phenotype at 30°C, and fully arrest at 37°C. Moreover, mutant cells in the SK1 background fail to grow normally even at the permissive temperature [16, 17]. Finally, Rrp6's important role in yeast is mirrored in mammals by our own results, and other recent reports showing that EXOSC10 is essential for cell division in mouse and human [19, 20, 21].

### *Rrp6 deletion and 5-FU treatment elicit an overlapping and conserved response.*

It was proposed earlier that 5-FU's toxic effect in yeast is based on DNA and RNA dependent mechanisms, but that the RNA pathway was dominant [5, 13]. A tiling array-based RNA profiling study of the 5-FU response in the fission yeast *S. pombe*, in combination with genetic analyses, provided additional evidence for 5-FU's effect on RNA processing; it is noteworthy that transcripts encoded by genes conserved between the distantly related budding- and fission yeasts, such as *SNU66*, *PRP8* and *ESF2*, respond to the drug (S6 Fig) [62]. Elegant work using pre-mRNA splicing reporter genes demonstrated that mRNA processing is impaired in a similar fashion when cells are treated with 5-FU or lack Rrp6, and that overexpressing *RRP6* reverses the effect of the drug on reporter gene expression [63]. Taken together, these data are consistent with the idea that an important, and at least partially conserved element of 5-FU's cytotoxic effect is mediated via altered RNA processing.

### *5-FU destabilizes transcription factors critical for cell division*

While mRNAs encoded by many cell cycle regulated genes showed the expected fluctuations after release into rich growth medium (see clusters C4 and C5 in Fig 3B), we found that some of the genes involved specifically in cell cycle progression (*AMNI*, *SIC1*, *PCL9*) and cell separation (*EGT2* and, among others, the *DSE* genes) failed to be induced to normal levels. A literature and database survey of activators targeting the promoters of protein-coding 5-FU target genes identified the paralogous transcriptional activators Swi5 and Ace2 that are known to be extensively regulated at the level of protein stability and localisation; reviewed

in [64]. Our data point to simultaneous down-regulation of Swi5 and Ace2 as a mechanism for 5-FU's ability to prevent the induction of certain cell cycle regulated genes. We did not expect to find that 5-FU affects regulatory proteins, however, the result is consistent with the earlier finding that 5-FU decreases the levels of mammalian transcription factors important for cell proliferation, such as ATF2, IRF2, and REL [65].

It is not clear why Swi5 is down-regulated by 5-FU but not *rrp6Δ*. It could be that the drug elicits a combination of effects such as decreased mRNA synthesis and protein instability caused by altered activity of cell cycle regulated proteases [56]; for review, see [64].

#### *Do lncRNA-mediated effects contribute to 5-FU's toxicity?*

The simplest explanation, which has been proposed before, is that modified RNAs are suboptimal substrates for Rrp6, and therefore tend to accumulate in 5-FU treated cells [6, 7]. This is consistent with our finding that lncRNAs, which increase when cells are exposed to 5-FU, strongly accumulate in cells lacking Rrp6. However, 5-FU might also up-regulate lncRNAs by negatively affecting conserved components of the RNA degradation machinery, such as EXOSC10/Rrp6, given that the gene is associated with an overlapping antisense lncRNA.

dsRNAs were proposed to play a major role in the regulation of bacterial gene expression [66]. In yeast, dsRNAs formed by s/a transcripts are involved in controlling the stability of mRNAs and XUT-type regulatory lncRNAs [58, 67]. In mammals, a dsRNA plays an important role during cell cycle progression by binding regulatory proteins [33]. In addition, we recently reported evidence that developmental stage-specific formation of dsRNAs is negatively correlated with protein levels [41]. We noticed that, in spite of the fact that *SWI5* and *ACE2* are paralogs, only the latter is associated with clearly expressed (and annotated) antisense lncRNAs (CUT259 and SUT2107). Both of these antisense transcripts accumulate in drug-treated and mutant cells and therefore recruit *ACE2* sense mRNA into a dsRNA form that is stable *in vivo*. This could prevent the mRNA from being transported into the cytoplasm or it could inhibit Ace2 protein translation by hindering the mRNA's association with ribosomes. In this context we note that, contrary to Swi5, Ace2 protein levels are down-regulated during the initial phases of cell cycle progression by both 5-FU and *rrp6Δ*, as one would expect if accumulating lncRNAs were to be involved.

*Are yeast 5-FU/Rrp6-targeted lncRNAs relevant for human cancer?*

The function of EXOSC10/Rrp6 as a subunit of the RNA exosome has evolved over time and so have its set of substrates or activities [68, 69, 70]. For example, it has been proposed that mammalian EXOSC10 may be more efficient at processing structured RNAs than yeast Rrp6 [71]. These differences notwithstanding, our data highlight the regulatory potential of 5-FU/Rrp6 responsive lncRNAs/dsRNAs in eukaryotes. This is a promising lead, especially in light of recent evidence that human EXOSC10 is an unfavorable prognostic marker for liver cancer [72].

Our yeast data provide global evidence that lncRNAs of all currently known major types accumulate in 5-FU treated cells, including antisense RNAs that form dsRNAs with overlapping sense mRNAs. Furthermore, we find again that dsRNAs are associated with decreased protein levels (see Fig 8; [41]). Is further experimental work in yeast, such as preventing the expression of specific antisense lncRNAs by deleting promoter regions, critical? Perhaps not. The evolutionary distance between yeast and human is large, and individual lncRNAs are not conserved at the sequence level from yeast to mammals. Consequently, a future priority is to identify, and further analyse EXOSC10/Rrp6-dependent human lncRNAs that respond to 5-FU using RNA-Sequencing and CRISPR/dCas9 based methods [73]. This approach will likely yield useful results, since earlier microarray-based work showed that 5-FU alters microRNA and lncRNA expression profiles in cultured breast- and colon cancer cells [74, 75, 76]. Critically, genome-wide mammalian RNA interactome data are now available [34, 35, 36, 77]. They should prove extremely useful in elaborating hypotheses about the roles of mammalian lncRNAs, and the transcripts they form dsRNAs with, in the etiology and progression of cancer.

Recent advances in tumor genomics and personalized medicine have helped conceptualize promising new approaches, such as immunotherapy in cancer treatment that complement chemotherapies today, and will perhaps replace them in the future [78, 79]. However, gaining insight into novel molecular mechanisms underlying the toxicity of 5-FU that has been used for decades, will likely continue to be important in the foreseeable future [80]. The yeast work reported here provides initial evidence in a simple eukaryotic model organism for roles of lncRNAs and dsRNAs in mediating 5-FU's toxicity. Such a mechanism would open up novel therapeutical targets that may help improve the outcome of established anti-cancer treatments.

## Materials and Methods

### *Strain backgrounds*

The strains employed for RNA profiling work were JHY222 *MATa/α* (derived from the standard yeast strain S288C) and JHY222 *MATa/α rrp6* that were described in [16], Table 1. In the JHY222 background, contrary to SK1, deletion of *RRP6* is not lethal under laboratory conditions, but causes a growth phenotype and impairs meiosis and spore formation [16, 17].

### *5-fluorouracil dose optimization*

Cells were cultured in YPD over-night and diluted to  $2 \times 10^6$  cells/ml in fresh YPD. 10 $\mu$ l of a suspension from serially diluted yeast cells (10-fold to  $10^4$ -fold) was deposited on standard YPD plates containing 20 $\mu$ g/ml 5-FU and cultured for 48 hours at 30°C or 37°C. Pictures were taken using the GE Imager system (General Electric, USA) at 48h.

### *Culture conditions and 5-fluorouracil treatment*

Wild type and *rrp6* mutant cells from fresh plates were inoculated in standard growth medium (YPD), and cultured over night at 30°C in a rotatory shaker (New Brunswick) for 19 hours. The cells were spun down and washed with pre-warmed water twice before they were re-inoculated into parallel cultures with growth medium (wild type and *rrp6Δ*) and YPD containing 20 $\mu$ g/ml 5-FU (wild type) at a concentration of  $2 \times 10^6$  cells/ml. Samples were harvested immediately after inoculation and then every 20 minutes until three hours, spun down, frozen in liquid nitrogen and stored at -80°C for further analysis. An aliquot was fixed in ethanol and used to monitor DNA synthesis.

### *Fluorescence activated cell sorting (FACS) analysis*

To monitor the progression of diploid cells through mitotic S-phase,  $10^7$  cells were resuspended, fixed with 70% ethanol and sonicated using a Branson Sonifier 250 for 10 seconds at 20% power, before they were analysed as described in [16].

### *Genome-wide strand-specific RNA-Sequencing library preparation*

For each condition, we isolated total RNA from biological duplicates using the hot phenol method as described [81]. Ribosomal RNA was removed using the RiboMinus Eukaryote System v2 kit (ThermoFisher, France). We next prepared strand-specific RNA-Sequencing libraries starting from 125 ng of rRNA-depleted RNA using the TruSeq Stranded Total RNA Sample Preparation kit (Illumina, USA), as described [58]. Before carrying out stranded RNA-Sequencing, we controlled total RNA and cDNA quality using the RNA 6000 Pico kit, and High Sensitivity DNA chips, respectively, and a BioAnalyzer 2100 (Agilent, France).

### *RNA-Sequencing data production and processing*

We employed a HiSeq2500 sequencer (Illumina) to carry out paired-end sequencing of the libraries (2 x 50 nucleotides). Reads were mapped onto the *S. cerevisiae* S288C reference genome using TopHat version 2.0.6 [82]. Parameters were set at a tolerance of three mismatches and a maximum intron size of 2 kb [30, 58]. Further bioinformatics analyses were carried out with uniquely mapped reads. Tag densities for genes, dubious ORFs, sn(o)RNAs, tRNAs and several classes of lncRNAs (including CUTs, SUTs, rsSUTs, XUTs and MUTs) were obtained after normalization over the total number of reads that uniquely mapped to ORFs. Mapping statistics are summarized in S1 Table.

### *RNA profiling data analysis*

To identify differentially expressed transcripts, we focused on transcripts with  $\geq 1$  count per million (CPM) in at least four out of 12 experiments. Such a filtering step is recommended for microarray and RNA-Sequencing based profiling experiments, since genes with very low counts across libraries do not provide robust evidence for differential expression [83, 84]. The library size was recomputed after this filtering step and counts were normalized using the TMM approach [85], and the differential analysis was performed with EdgeR [86, 87]. The tested contrasts were either focused on the condition and the time point. The Benjamini and Hochberg correction for multiple testing was applied and transcripts with a corrected  $p$ -value  $< 0.05$  were selected as being significantly differentially expressed [88]. The analysis was performed using the EdgeR Bioconductor package [89].

### *Clustering and Gene Ontology Term enrichment*

Different strategies were employed to determine the optimal clustering method and its parameters. We tested from  $k=4$  to  $k=20$  using  $k$ -means [86], and Partitioning Around Medoids (PAM) [87]. We generated silhouette plots of each step to assess the clustering methods' efficiency. This, together with visual inspection of the heatmaps enabled us to identify  $k$ -means,  $k=5$  as the best approach. Gene Ontology (GO) Biological Process term enrichment was computed using GoTermFinder whereby GO terms with adjusted  $p$ -values below 0.01 were considered as significant [90]. For clusters with enriched GO terms, ReviGO was then used to cluster the significant GO terms based on their semantic similarity (default parameters) [91].

### *Raw- and processed data availability*

Raw data files were uploaded to the NCBI's GeneOmnibus (GEO) repository and are available under the accession number GSE77511 [92]. The RNA-Sequencing data for all replicates and averaged duplicates are also conveniently accessible online in a searchable viewer at <http://vm-gb.curie.fr/mprimig/5FU>.

### *Genome-wide detection of dsRNAs*

Genome-wide data on dsRNA formation in haploid wild type and *xrn1* mutant cells were previously published [58]. The RNA-Sequencing data are available online at [http://vm-gb.curie.fr/mw2/small\\_RNA-seq/](http://vm-gb.curie.fr/mw2/small_RNA-seq/).

### *Immunological detection of dsRNA*

An RNA immunoprecipitation (RIP) assay was performed using the dsRNA-specific J2 antibody [59]. RNA was extracted with cold phenol. 35 $\mu$ g RNA was incubated with 3.5 $\mu$ g dsRNA-specific J2 antibody (Scicons, France) at 4°C overnight. Dynabeads were prepared according to the manufacturer's instructions and incubated with the antibody-RNA mixture on a rotator at 4°C for 2h. RNA was washed, eluted and treated with DNaseI (Ambion, France). Finally, precipitated RNA was reverse transcribed using the High Capacity cDNA Reverse Transcription kit (ThermoScientific, USA) and amplified using Taq Polymerase (Qiagen, France) at 60°C for 30 cycles using the primer pair for *ACE2* (Table 2).

### *RT-PCR assays*

Strand-specific RT-PCR assays were carried out using the SuperScript II kit (ThermoScientific, USA). 2µg of RNA was treated with DNaseI (Ambion, USA), and reverse transcribed with reverse primers (shown in Table 2), in the presence of Actinomycin D to prevent second strand synthesis. After reverse transcription, the samples were treated with RNase H (TEBU BIO, France) to remove the RNA template. cDNA was PCR amplified using Taq polymerase (Qiagen, USA) at 60 °C for 28 cycles. RT-PCR products were separated on 2% agarose gels and photographed using the Gel-doc XRS image system (Bio-Rad, USA). Primers used for RT-PCR were designed using NCBI Primer-Blast program. Standard RT-PCR assays were performed as published [93]. Briefly, RNA was extracted using the hot phenol method, followed by DNase I treatment with TURBO DNA-free Kit (Ambion, USA). RT-PCR reactions were carried out using 2 µg of RNA reverse transcribed with Reverse Transcriptase (High Capacity cDNA Reverse Transcription kit; Life Technologies, USA) and amplified using Taq Polymerase (Qiagen, France) at 60°C for 28 cycles. RT-PCR products were separated on 2% agarose gels and photographed using the Gel-doc XRS image system (Bio-Rad, USA). Primers used for RT-PCR were designed using Primer3 (simgene.com/Primer3; Table 2).

### *Western blotting*

Protein extracts were prepared from samples of cells cultured in rich medium (YPD), pre-sporulation medium (YPA), and sporulation medium (SPII) as described in [94]. 20µg of total protein extract was separated by SDS-PAGE gel electrophoresis (Bio-Rad, USA) and transferred to ImmobilonPSQ membranes (Millipore, France) using an electro-blotter (TE77X; Hoefer, USA) set at 50mA for 2.5h. The membrane was blocked with 5% milk powder at room temperature for 1h, incubated with anti-Swi5 and anti-Ace2 antibodies (1:1000) overnight at 4°C. A secondary anti-rabbit antibody (1:10000) was incubated at room temperature for 1h. An anti-Pgk1 antibody (1:10000) was incubated at room temperature for 1h before the signals were revealed using an anti-mouse secondary antibody (1:10000) incubated at room temperature for 1h. Protein signals were revealed using the ECL-Plus Chemiluminescence kit (GE Healthcare, USA) and the ChemiDoc XRS system (Bio-Rad, USA). Band intensities were quantified with the ImageQuant TL 7.0 software at standard settings (GE Healthcare, USA).

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## **Declaration of interest**

The authors declare no interest.

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## Figure legends

**Fig 1. Experimental approach.** A flow chart shows the color coded wild type (green) and mutant (red) strains that were employed in the RNA profiling experiment including cell culturing, data production and analysis and validation steps.

**Fig 2. Monitoring cell growth and DNA replication.** (A) Images show the outcome of a plate growth assay after 48 hours at the temperatures indicated at the bottom on rich medium (YPD) at permissive (30°C) and restrictive temperatures (37°C) for wild type (WT in green) and mutant (*rrp6* in red) cells in the strain indicated to the left. The drug concentration is given to the right. Grey triangles at the top indicate increasing dilution of the cell suspension. (B) The output of a FACS analysis is given. The time points are given in minutes and are shown to the left; the samples that were further analysed are shown in red. Color-coded untreated wild type (WT in green), treated wild type (WT 20µg/ml 5-FU in purple) and mutant (*rrp6* in red) cell samples are shown at the top. DNA content and strain background are given at the bottom and to the right, respectively.

**Fig 3. RNA-Sequencing analysis strategy and output.** (A) A flow chart outlines the RNA-Seq data analysis strategy. (B) A color-coded heatmap (red is high and blue is low) shows the expression patterns for protein-coding RNAs (mRNAs) grouped into five clusters (C1-C5) given to the left for sample duplicates from wild type (WT), treated wild type (5-FU) and mutant (*rrp6Δ*) cells as shown at the top. The time points are indicated in minutes. The clusters are biologically classified into genes that respond to treatment and gene deletion (5-FU/*rrp6Δ*), gene deletion (*rrp6Δ*) and cell cycle peaks (G1/S and G2/M). [A color scale is shown at the bottom.](#) (C) Color-coded histograms are shown for G1/S and G2/M marker genes as given at the top. The sequence (x-axis) is plotted against expression signals (y-axis) for wild type cells (WT40, WT100, shades of green), treated cells (5-FU40, 50-FU100, shades of blue) and mutant cells (*rrp6Δ*40, *rrp6Δ*100, shades of red) at 40 and 100 minutes after release into fresh medium. The data range set in IGV is indicated.

**Fig 4. Expression data for cell cycle genes.** (A) A histogram of RNA-Sequencing data is shown for *AMN1* like in Fig 3C. (B) An RT-PCR assay is shown (top) and band intensities are quantified (bottom) for *AMN1*. *ACT1* was used as a control. Samples in lanes 1-6 are indicated at the top and bottom. The strain background is given to the left. (C) Expression data are shown as histograms for selected cell cycle genes as indicated at the top, like in Fig 3C.

**Fig 5. *SWI5* RNA and protein analysis.** (A) A histogram is shown for *SWI5* like in Fig 3C. (B) A Western blot assay is shown for Swi5 in strain backgrounds indicated at the bottom. Pgk1 was used as a loading control. Non-specific bands are indicated with an asterisk to the right. Lanes 1 and 2 contain samples from wild type (WT) and *swi5* deletion (*swi5* $\Delta$ ) strains. Lanes 3-8 contain wild type cells (WT40min, WT100min), treated cells (5-FU40min, 5-FU100min) and mutant cells (*rrp6* $\Delta$ 40min, *rrp6* $\Delta$ 100min) 40 and 100 minutes after release into fresh medium.

**Fig 6. RNA profiling of lncRNAs.** A color coded heatmap is shown for lncRNAs like in Fig 3B. Clusters (C1-C5) and response classes are given to the left and right. Sample duplicates are indicated at the top. A color scale is shown at the bottom.

**Fig 7. Expression data for antisense lncRNAs.** (A) A color coded heatmap for antisense lncRNAs (AS lncRNAs) is shown as in Fig. 6. (B) Color-coded genomic heatmaps are shown for mRNAs in blue, and lncRNAs in light grey (SUTs), red (XUTs) and green (CUTs). Chromosome numbers and genome coordinates are given in the middle. Red boxes delineate selected lncRNAs. Wild type (WT), treated (WT 5-FU) and mutant (*rrp6*) samples, and the time points in minutes are shown to the right. Top (+) and bottom (-) DNA strands and a scale are given to the right. Images were generated using VING software [95]. (C) A bar diagram shows quantified RNA-Sequencing expression signals for SUT597. The expression signal (y-axis) is plotted against the samples (x-axis) as shown at the bottom. Time points are given for wild type (WT), drug-treated (5-FU) and mutant (*rrp6*) samples. Data are presented as mean values +/- SEM (calculated from two biological replicates used in the RNA-Sequencing analysis). (D) An RT-PCR assay is shown for antisense lncRNAs and overlapping sense mRNAs as indicated to the right. Wild type (WT), treated (WT 5-FU) and

mutant (*rrp6*) samples, and the time points in minutes are given at the top. *ACT1* was used as a loading control. (E) A color-coded graph plots the 19-23 base small RNA read densities (tag/nucleotide, y-axis) against top (+) and bottom (-) strands of the genome (y-axis). Protein-coding and non-coding genes are represented by arrows for which the color code is like in panel B. Data are shown in black for a wild type strain (WT, dotted line), and a strain expressing Dicer/Argonaute (RNAi), and in red for a mutant strain (*xrn1*, dotted line), and a mutant strain expressing Dicer/Argonaute (*xrn1* RNAi). Chromosome numbers and genome coordinates are given at the top and bottom, respectively. The images were generated using the VING software and previously published small RNA-Sequencing data [58, 95]. To view a larger region covering the loci and a graphical representation of the signals showing wild type and engineered RNAi strain separately, see [http://vm-gb.curie.fr/mw2/small\\_RNA-seq/](http://vm-gb.curie.fr/mw2/small_RNA-seq/).

**Fig 8. Sense/antisense RNA and protein expression data for *ACE2/SUT2107/CUT259*.**

(A) A genomic expression heatmap for the region containing *ACE2/SUT2107/CUT259* is shown like in Fig 7B. (B) A bar diagram shows quantified RNA-Sequencing expression signals for *ACE2*, *CUT259* and *SUT2107* as in Fig 7D. (C) An RT-PCR assay is shown for *ACE2* and *SUT2107* like in Fig 7E. A non-specific band is marked with an asterisk. *ACT1* was used as a loading control. (D) A Western blot assay is shown for *Ace2* like in Fig 5B. A non-specific band is marked with an asterisk. *Pgk1* was used as a loading control. Samples are given at the bottom and at the top. Band intensities are shown in a histogram where samples (x-axis) are plotted against relative intensity units (y-axis). (E) A color-coded graph plots the 19-23 base small RNA read densities (tag/nucleotide, y-axis) against top (+) and bottom (-) strands of the genome (y-axis) like in Fig 7 F. (F) An immunoprecipitation (IP) assay is shown using the dsRNA specific J2 antibody for the s/a pair given at the bottom. RNA samples harvested 100 minutes after release into fresh medium from wild type (WT), drug-treated (5-FU) and mutant (*rrp6Δ*) cells were assayed in the presence (+) or absence (-) of J2 antibody as shown at the top. A PCR product revealing precipitated RNA is marked to the right (dsRNA).

**S1 Fig. Global expression patterns.** (A) A scatterplot matrix is shown for pair-wise comparisons of averaged levels calculated from duplicate samples for all annotated

transcripts. (B) Color coded graphs to the left plot the ratios of log<sub>2</sub>-transformed expression signals for mutant (*rrp6*) and wild type (WT) cells (left), and treated (5-FU) and untreated (WT) cells (right), at the 40-minute time point (x-axis) against the tag density (y-axis). The legend indicates the color code for different classes of lncRNAs. Box plots to the right show the mutant over wild type signal ratio distributions (y-axis) for the given transcript types (x-axis).

**S2 Fig. Gene Ontology (GO) term enrichment.** (A-C) Enrichment data are shown as ReviGO plots for cluster C3, C4 and C5.

**S3 Fig. dsRNA formation *in vivo* in haploid cells.** A graphical display from the online viewer at [http://vm-gb.curie.fr/mw2/small\\_RNA-seq/](http://vm-gb.curie.fr/mw2/small_RNA-seq/) is given for the genes delineated by vertical red lines. Bar diagrams show log<sub>2</sub> tag densities representing reads (y-axis) for 15 kb regions (x-axis). Genome coordinates are given at the bottom and the chromosome number is shown at the top. Samples from a wild type (WT) control and a genetically altered strain expressing Dicer and Argonaute (RNAi) are given at the top. Top (+) and bottom (-) strands are indicated to the right. Transcripts are color-coded in blue (mRNAs), red (XUTs), grey (SUTs), green (CUTs) and black (tRNAs). Arrows indicate the transcriptional direction.

**S4 Fig. *MTR3*/*CUT158* expression and dsRNA formation.** (A) A heatmap is shown for the genomic region containing the *s/a* locus like in Fig. 7B. (B-C) RNA-Sequencing and RT-PCR data are plotted like in Fig. 7C and D. (D) dsRNA profiling data are shown like in Fig. 7E.

**S5 Fig. *GAPI*/*SUT1912* expression and dsRNA formation in mitosis and meiosis.** (A) A color-coded heatmap is shown for the genomic region containing the *s/a* locus. Red and blue represent high and low log<sub>2</sub>-transformed values, respectively. Each line shows a sample from cells cultured in rich medium with glucose (YPD), presporulation medium (YPA) or sporulation medium (SPII 1-12h) as given to the left. Each column represent the values for an oligonucleotide probe present on a yeast tiling array. Chromosome numbers, top (+) and bottom (-) strand are indicated. Horizontal black lines represent DNA strands. ORFs are given in blue, lncRNAs in mauve. Genome coordinates are shown. (B) dsRNA-profiling data

are shown for the genomic region like in S3 Fig. Blue dotted arrows represent dubious ORFs. (C) A J2 RIP assay for *GAPI/SUT1912* is shown. Lane 1 contains molecular weight markers (MW). Samples from cells cultured in rich medium and sporulation medium were precipitated as given at the top.

**S6 Fig. 5-FU response of genes conserved between budding- and fission yeast.** Bar diagrams of RNA-Sequencing data are shown for three conserved genes involved in RNA processing, like in Fig 7A.

**S1 Table. RNA-Sequencing mapping statistics.**

**S2 Table. Cluster distribution for mRNA, lncRNAs and all transcripts.**

**S3 Table. GO term enrichment statistics for mRNA clusters.**

**S4 Table: Cluster distribution for sense/antisense loci**

**Table 1. Yeast strains**

Strain number	Genotype	Reference and source
MPY1	JHY222 <i>MATa/α</i>	[16]
MPY392	JHY222 <i>MATa/α rrp6/rrp6</i>	[16]
MPY827	BY4743 <i>MATa/α</i>	EUROSCARF
MPY828	BY4743 <i>MATa/α swi5/swi5</i>	J. Horecka
MPY830	BY4743 <i>MATa/α ace2/ace2</i>	EUROSCARF

**Table 2. Oligonucleotides for RT-PCR assays.**

Target	Forward primer	Reverse primer
<i>ACE2</i>	5'-TGCCCATGCGGAAAGAGATT-3'	5'-TCGCAGCTGTTCCCTCCATTT-3'
<i>ACT1</i>	5'-CTGCCGGTATTGACCAAAC-3'	5'-AGATGGACCACTTTCGTCGT-3'
<i>AMN1</i>	5'-AAGGAGAAGAAGCAAGGGGC-3'	5'-GCAAACCCACAGTTTCCAC-3'
<i>CUT158</i>	5'-AGGACACCAAACATCGCTCA-3'	5'-TTTGCTTIGCGAATCGTCGT-3'
<i>MNL1</i>	5'-GCAGGAAGTGAAGGAGCTGT-3'	5'-TGCCATATTCTGGGCGAGAC-3'
<i>MTR3</i>	5'-GCTAGCCTTAGCTGATGCCG-3'	5'-GCAACGATCAAGGCATTCCA-3'
<i>SUT597</i>	5'-TGTCTACTCGTCGCACTTGG-3'	5'-GTAGCCTCTTCGTCTGGTGG-3'
<i>SUT2107</i>	5'-CTGCAATGACACCCACAAA-3'	5'-CCTCGTGCTTGTATGACTGC-3'
<i>SWI5</i>	5'-CGCTGGGACCACTTTCTGAT-3'	5'-TCGCGCAACTCTTGTGTTG-3'
<i>GAP1</i>	5'-TTGTTGCCGCCTCCAAAAG-3'	5'-CCCCAGTAGGAACCCCAAAC-3'

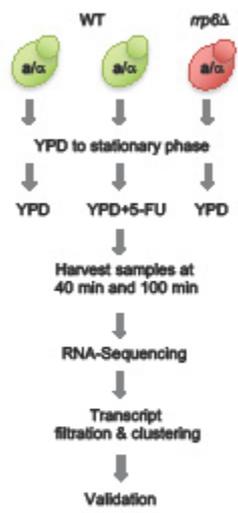


Figure 1

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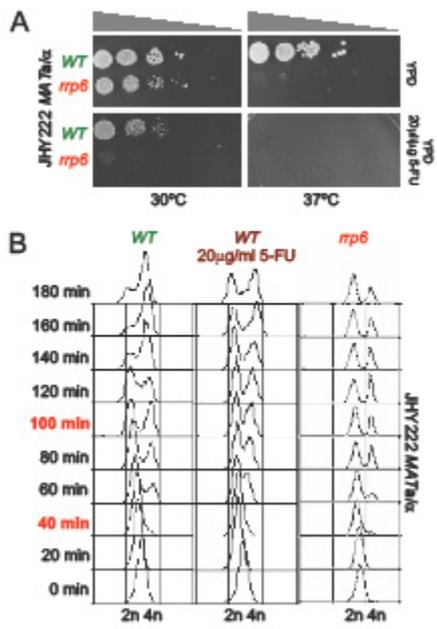


Figure 2

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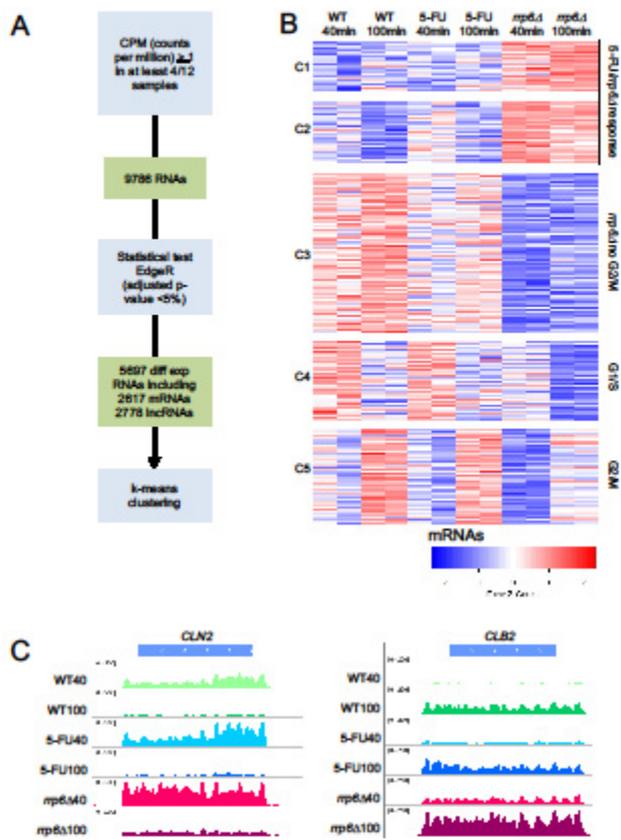


Figure 3

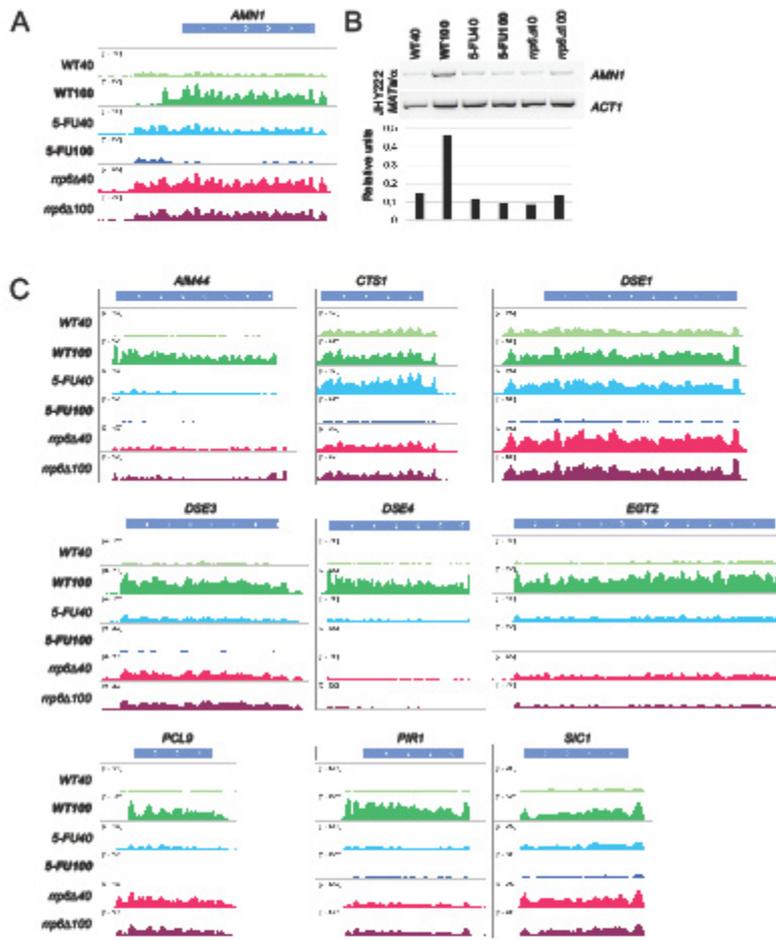


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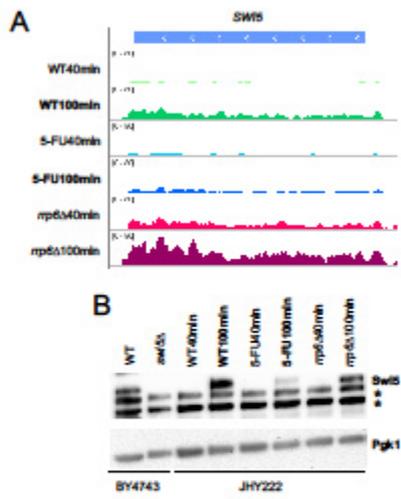


Figure 5

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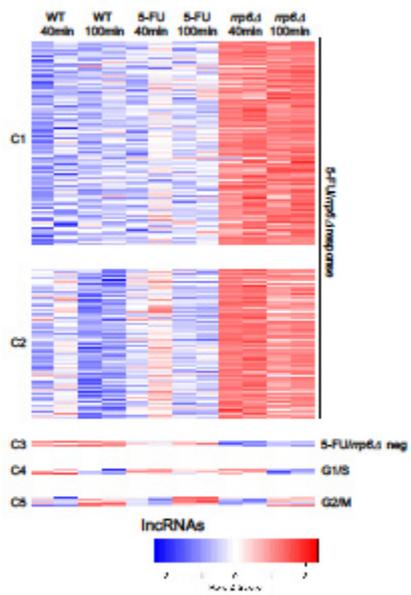


Figure 6

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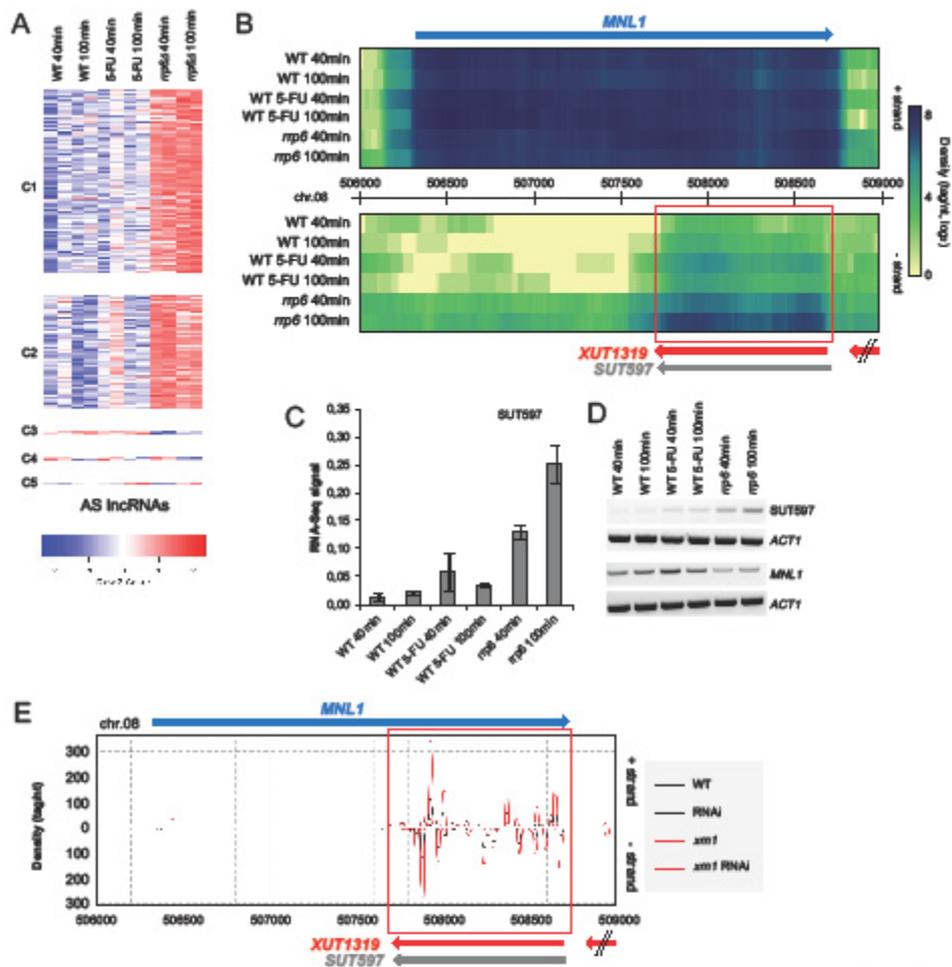


Figure 7

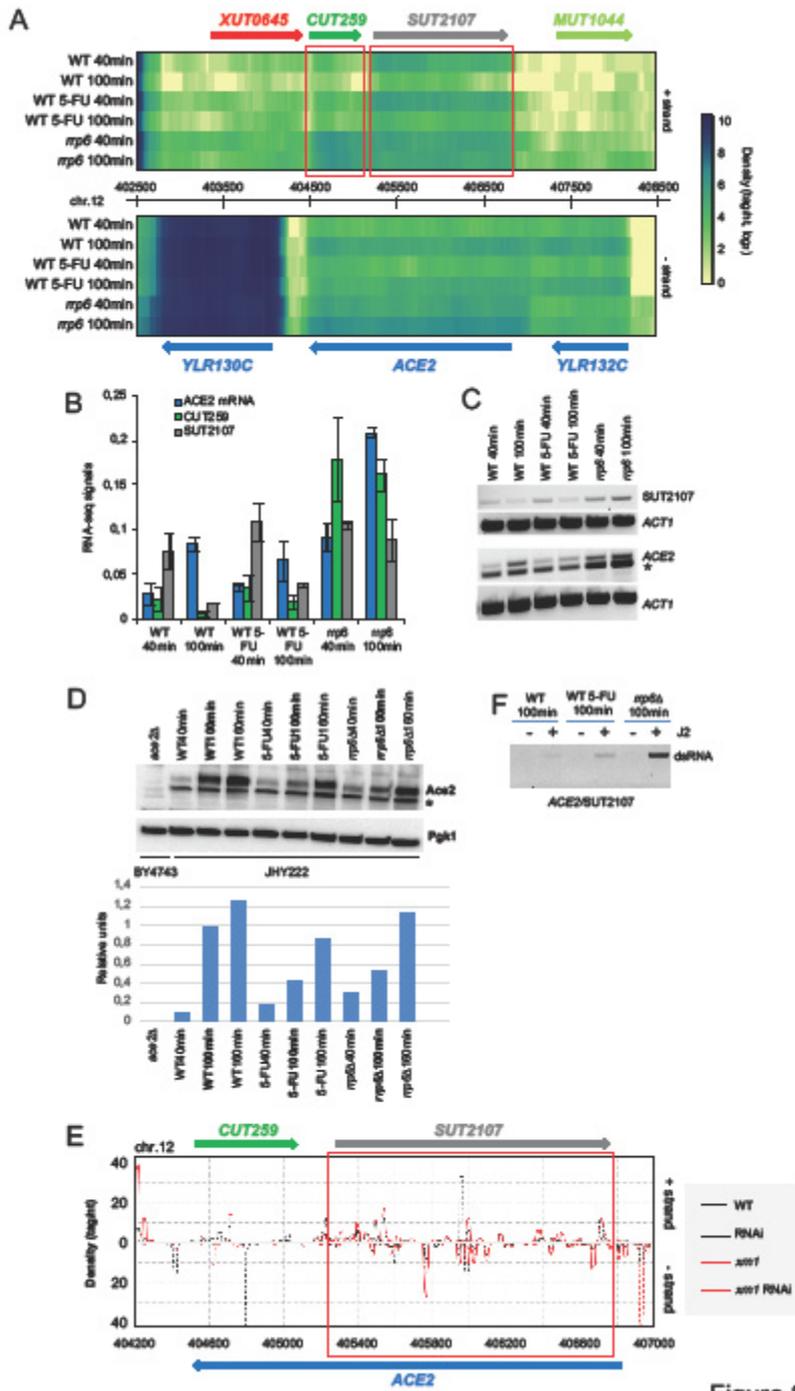
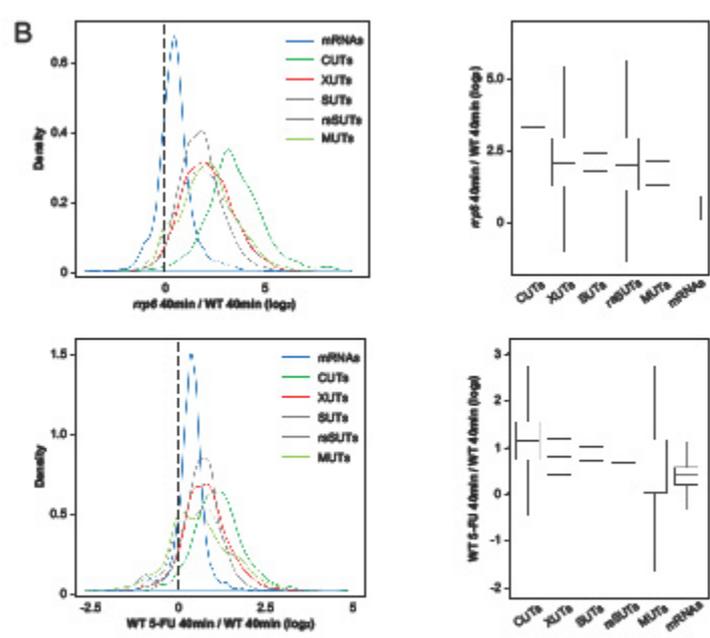
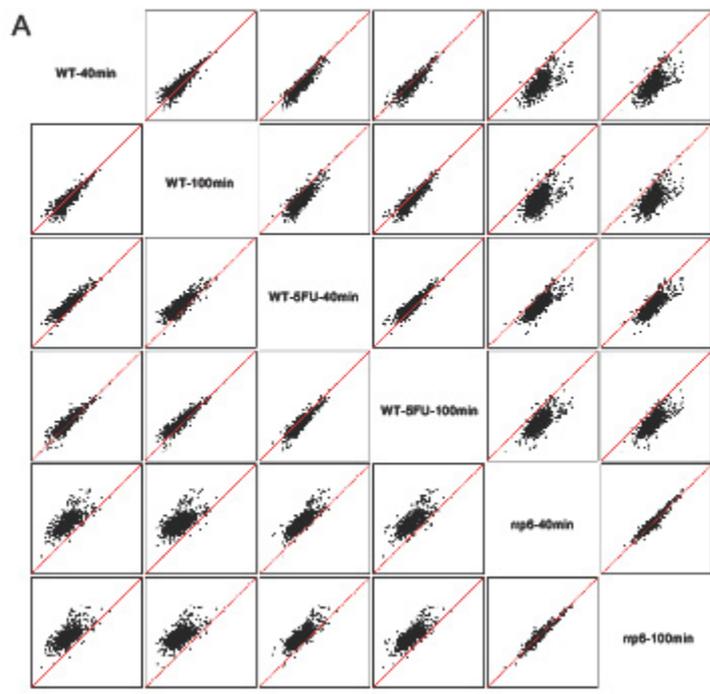
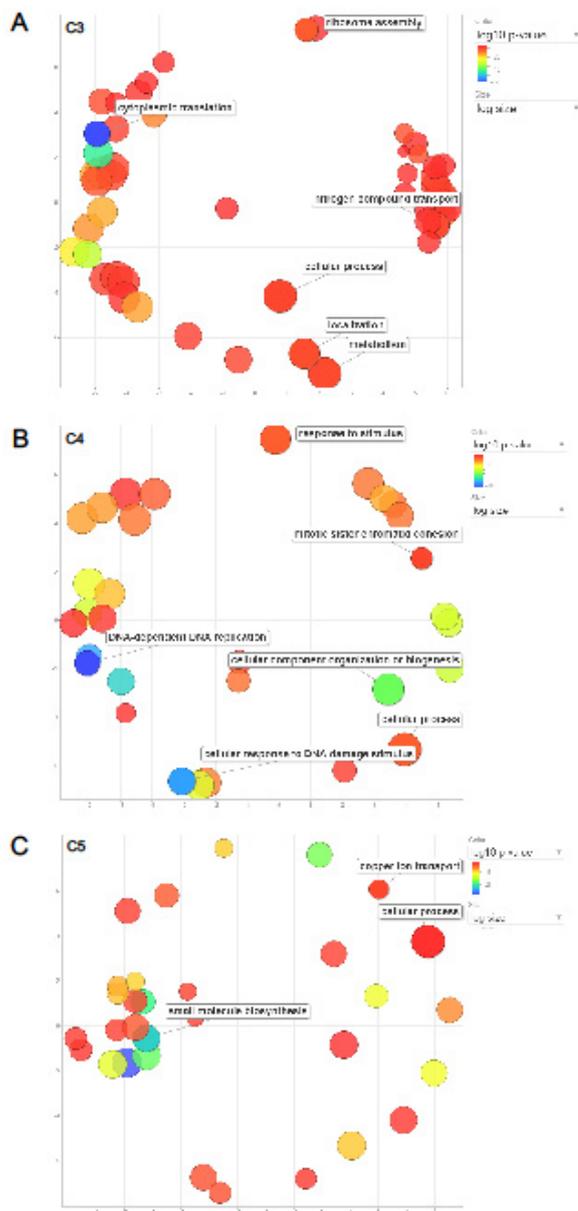


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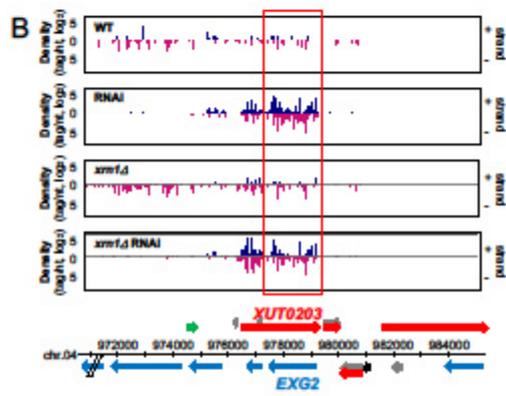
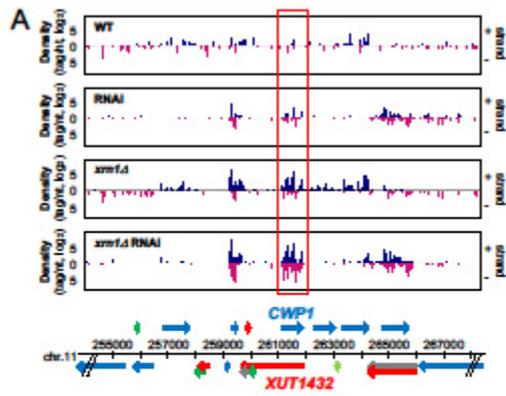
S1 Figure



S2 Figure

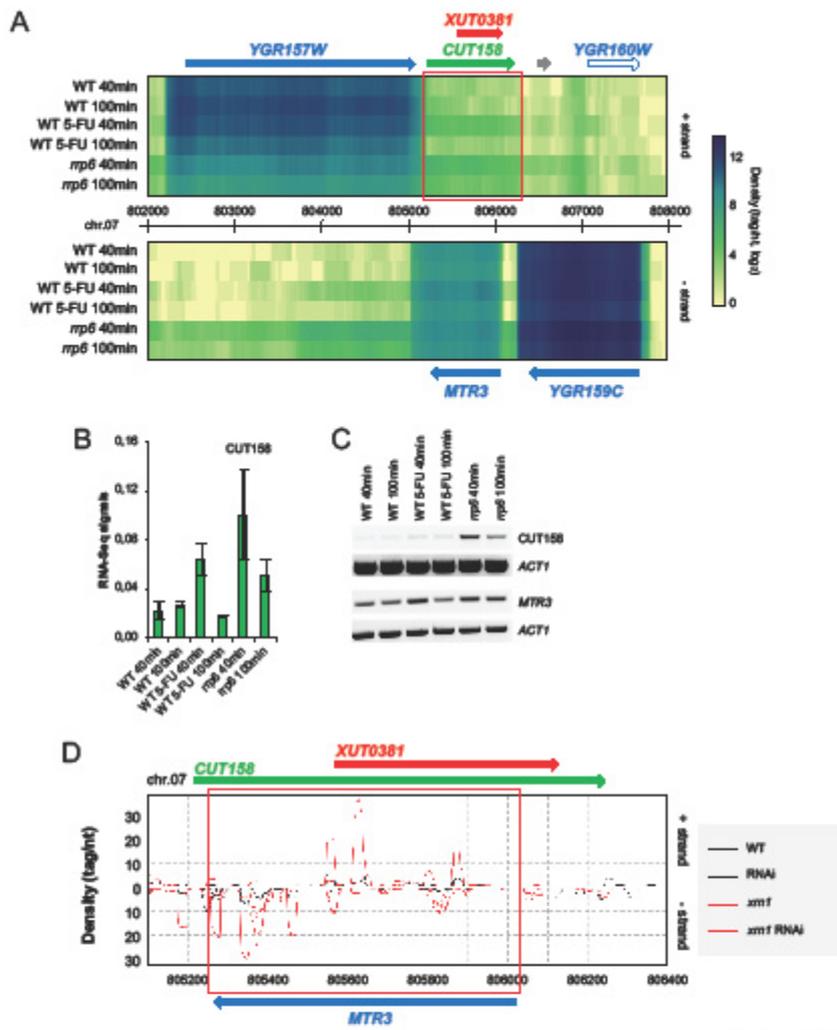
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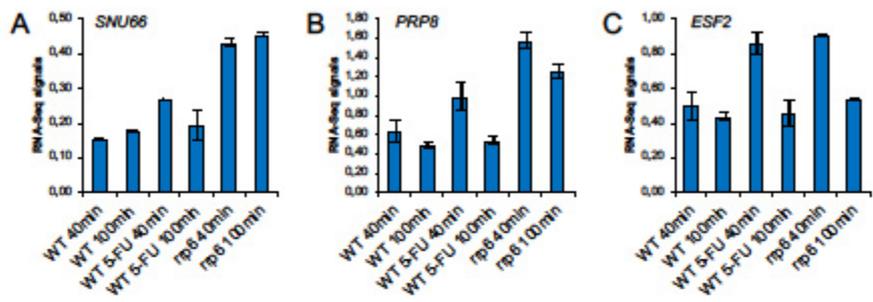
S3 Figure

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S4 Figure





S6 Figure

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