



Developmental Dynamics of Long Noncoding RNA Expression during Sexual Fruiting Body Formation in *Fusarium* graminearum

Wonyong Kim,^a Cristina Miguel-Rojas,^a Jie Wang,^a Jeffrey P. Townsend,^{b,c} Frances Traila,^d

- ^aDepartment of Plant Biology, Michigan State University, East Lansing, Michigan, USA
- ^bDepartment of Biostatistics, Yale University, New Haven, Connecticut, USA
- ^cDepartment of Ecology and Evolutionary Biology, Yale University, New Haven, Connecticut, USA
- ^dDepartment of Plant, Soil and Microbial Sciences, Michigan State University, East Lansing, Michigan, USA

ABSTRACT Long noncoding RNA (IncRNA) plays important roles in sexual development in eukaryotes. In filamentous fungi, however, little is known about the expression and roles of IncRNAs during fruiting body formation. By profiling developmental transcriptomes during the life cycle of the plant-pathogenic fungus Fusarium graminearum, we identified 547 IncRNAs whose expression was highly dynamic, with about 40% peaking at the meiotic stage. Many IncRNAs were found to be antisense to mRNAs, forming 300 sense-antisense pairs. Although small RNAs were produced from these overlapping loci, antisense IncRNAs appeared not to be involved in gene silencing pathways. Genome-wide analysis of small RNA clusters identified many silenced loci at the meiotic stage. However, we found transcriptionally active small RNA clusters, many of which were associated with IncRNAs. Also, we observed that many antisense IncRNAs and their respective sense transcripts were induced in parallel as the fruiting bodies matured. The nonsense-mediated decay (NMD) pathway is known to determine the fates of IncRNAs as well as mRNAs. Thus, we analyzed mutants defective in NMD and identified a subset of lncRNAs that were induced during sexual development but suppressed by NMD during vegetative growth. These results highlight the developmental stage-specific nature and functional potential of IncRNA expression in shaping the fungal fruiting bodies and provide fundamental resources for studying sexual stage-induced lncRNAs.

IMPORTANCE Fusarium graminearum is the causal agent of the head blight on our major staple crops, wheat and corn. The fruiting body formation on the host plants is indispensable for the disease cycle and epidemics. Long noncoding RNA (lncRNA) molecules are emerging as key regulatory components for sexual development in animals and plants. To date, however, there is a paucity of information on the roles of lncRNAs in fungal fruiting body formation. Here we characterized hundreds of lncRNAs that exhibited developmental stage-specific expression patterns during fruiting body formation. Also, we discovered that many lncRNAs were induced in parallel with their overlapping transcripts on the opposite DNA strand during sexual development. Finally, we found a subset of lncRNAs that were regulated by an RNA surveillance system during vegetative growth. This research provides fundamental genomic resources that will spur further investigations on lncRNAs that may play important roles in shaping fungal fruiting bodies.

KEYWORDS Fusarium graminearum, Xrn1 exonuclease, fruiting body, long noncoding RNAs, perithecia, sexual development

Received 16 June 2018 **Accepted** 5 July 2018 **Published** 14 August 2018

Citation Kim W, Miguel-Rojas C, Wang J, Townsend JP, Trail F. 2018. Developmental dynamics of long noncoding RNA expression during sexual fruiting body formation in Fusarium graminearum. mBio 9:e01292-18. https://doi.org/10.1128/mBio.01292-18.

Editor Antonio Di Pietro, Universidad de Córdoba

Copyright © 2018 Kim et al. This is an openaccess article distributed under the terms of the Creative Commons Attribution 4.0 International license.

Address correspondence to Frances Trail, trail@msu.edu.

enomes of eukaryotes—from simple yeast to animals—are pervasively transcribed lacksquare from noncoding intergenic regions and in antisense orientation from genic regions (1, 2). Long noncoding RNAs (IncRNAs) are loosely defined as noncoding transcripts longer than 200 nucleotides, which are mostly transcribed by RNA polymerase II and share common features with mRNAs other than protein-coding capacity (3). IncRNAs are versatile molecules that not only regulate gene expression, but also affect enzymatic activities and chromosome conformation (4, 5). Since the discovery of the XIST IncRNA required for X chromosome inactivation (6), the roles of IncRNAs in developmental processes such as embryogenesis and tissue differentiation have been extensively studied in animals along with the advent of transcriptome sequencing (RNA-seq) technologies (7, 8). Yet the full scope of the developmental roles of IncRNAs is far from understood.

In the highly divergent yeasts Saccharomyces cerevisiae (budding yeast) and Schizosaccharomyces pombe (fission yeast), the onset of sexual sporulation and the following meiotic divisions are tightly regulated by elaborate mechanisms involving IncRNAs (9). In budding yeast, a promoter-derived IncRNA suppresses the expression of IME1 (inducer of meiosis 1), the master regulator for the sexual sporulation, by inducing heterochromatin formation in the promoter region of IME1 during vegetative growth (10). In addition, the transcription of another lncRNA antisense to IME4 gene inhibits the expression of IME4 by antagonizing sense transcription (11, 12). Although there is no such conserved or analogous regulatory mechanism in fission yeast, IncRNAs also play diverse roles in sexual sporulation: for example, sequestering RNA elimination factors that repress meiotic gene expression (13-15) and contributing to homologous chromosome pairing (16). Despite the growing evidence of the regulatory roles in yeasts, information on IncRNA expression and function during fruiting body formation in filamentous fungi is scarce.

RNA quality control mechanisms are crucial for the regulation of lncRNA expression in budding yeast. The nuclear exosome is engaged in RNA processing and degradation of transcripts, including IncRNAs that are specifically expressed during sexual sporulation; the deletion of RRP6 encoding the exosome-associated exonuclease resulted in the accumulation of noncoding transcripts that otherwise remained silenced during vegetative growth (17-19). In human cells, promoter-derived transcripts were also ectopically expressed upon deletion of the exosome components, including the homologous RRP6 gene, suggesting the conserved role of the exosome for IncRNA expression in diverse eukaryotes (20). The nonsense-mediated decay (NMD) pathway is another quality control checkpoint for aberrant transcripts in the cytoplasm and recently emerged as a key player for fine-tuning of both coding and noncoding gene expression (21). A genome-wide survey of human IncRNA sequences showed that most IncRNAs harbor short open reading frames (ORFs) that would lead to activation of the NMD pathway (22). In fact, subsets of lncRNAs found in budding yeast, the model plant Arabidopsis, and animals are subject to degradation through the NMD pathway (23–25).

The cytoplasmic exonuclease Xrn1 is the final enzyme responsible for the degradation of decapped and deadenylated transcripts that have been recognized and processed by NMD components. The deletion of XRN1 in budding yeast also leads to the accumulation of more than a thousand cryptic noncoding transcripts termed "XUTs" (Xrn1-sensitive unstable transcripts), most of which are distinct from the noncoding transcripts that arise by exosome depletion (26). Many XUTs are antisense to annotated genes and seemed to have repressive roles in sense transcription by modulating chromatin status of the promoter regions (26).

It has been argued that organismal complexity is correlated with expression dynamics of noncoding transcripts (5, 27–29). In the kingdom Fungi, multicellular fruiting bodies have independently evolved at least twice in the diverging lineages (30, 31). Given the key regulatory roles of lncRNAs in sexual sporulation and morphological transition in yeasts (32, 33), IncRNAs may have exerted their roles in evolution of multicellularity and sexual development in filamentous fungi. Fusarium graminearum is a plant-pathogenic fungus infecting our staple crops, such as wheat and corn, and thus

mBio

has been a model for studying the developmental process of perithecia, the sexual fruiting bodies of the fungus, as well as other interesting aspects of biology, including host-pathogen interaction and mycotoxin production (34, 35). The fungus is probably the best organism for investigating the lncRNA catalog in fruiting body-forming fungi, as perithecia develop at sufficient synchronicity in culture media, enabling time-series transcriptome analyses with this microscopic organism (36-38). Also, the genome sequence assembly is complete, featuring a total of 4 chromosomes (39), and the genome has been annotated and curated—although it still lacks IncRNA annotations (40, 41). In addition, plentiful genetic resources have accumulated through large-scale functional studies of perithecium development (38, 42-47).

The goal of the present study was to characterize IncRNAs that are specifically expressed in the fungal fruiting body undergoing sexual development and to investigate their developmental stage-specific expression and regulation. We identified IncRNAs with a pipeline that constructs de novo transcript annotations by combining RNA-seq data from vegetative and sexual developmental stages and then removes those with detectable protein-coding potential or a monotonous expression profile. Hundreds of IncRNAs that exhibit dynamic expression patterns were found, thereby expanding the universe of genomes known to have significant noncoding roles in development—specifically, here, the multicellular development of fungal fruiting bodies.

RESULTS

Transcriptional reprogramming during perithecium formation. To obtain time course transcriptome data during the sexual development of F. graminearum, we sequenced samples from hyphae, strands of cells that make up vegetative stages of most of the fungi (S0), and from five successive sexual stages (S1 to S5 [Fig. 1A]) that capture key morphological transitions during the development of perithecia (48), defined at the formation of the following: S1, perithecial initials (hyphal curls that give rise to the perithecial tissues); S2, perithecial walls; S3, paraphyses (sterile cells supporting perithecia): S4. asci (saclike structures in which ascospores develop): and S5. ascospores (meiospores). A total of 480 million RNA-seg reads were generated from 18 samples (6 stages imes 3 replicates), and there were an average of 25 million mapped reads per sample (see Fig. A in Text S1 in the supplemental material). We validated our sampling scheme by perithecial morphology for 3 biological replicates, using the BLIND program (49), which determined the sequence of the developmental time course data without prior information other than gene expression data (see Fig. B in Text S1).

Differentially expressed (DE) genes between any two successive developmental stages (>4-fold at a 5% false-discovery rate [FDR]) were mostly unique compared to other pairwise stage comparisons (Fig. 1B). Overrepresented Gene Ontology database (GO) terms for the stage-specific DE genes reflected key biological processes during the morphological transitions (see Table A in Text S1). For example, the GO term "lipid metabolism" had the highest representation in the "S0 versus S1" comparison, although it was statistically nonsignificant (Fig. 1C). The accumulated lipids in hyphae and perithecial initials are vital for development of paraphyses and asci (50). Perithecia dramatically increased in size and became more rigid during S2 and S3, which is accompanied by GO terms related to carbohydrate metabolism (Fig. 1C). Finally, when the asci develop from the fertile layer during S3 and S4, meiosis-related genes were significantly enriched for GO terms, including "meiotic cell cycle process" and "ascospore formation" at a 5% FDR (Fig. 1C). It is noteworthy that most of the DE genes were upregulated in the later developmental stages (see Fig. C in Text S1), indicating that gene activation is the most common means of gene regulation during sexual development.

Identification of IncRNA in perithecia. To discover IncRNAs expressed during perithecial development, we adopted an established protocol for novel transcript identification (51), with some modifications (see Fig. D in Text S1). First, we constructed de novo transcript annotations (28,872 transcripts expressed from 20,459 genomic loci)

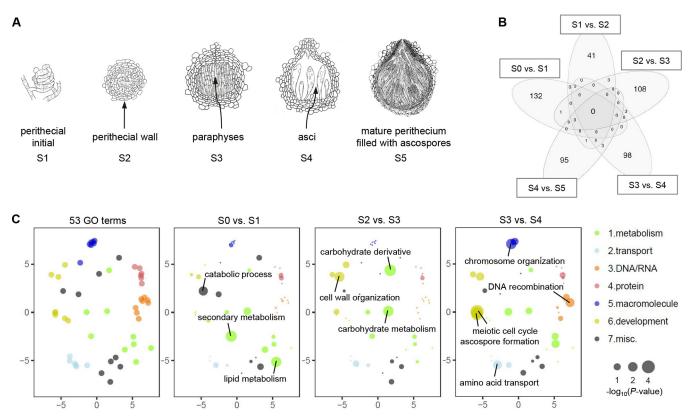


FIG 1 Transcriptome of F. graminearum perithecia. (A) Emergence of new tissues at the defined developmental stages during perithecial formation (S1 to S5 [not drawn to scale]). (B) Venn diagram showing the number of differentially expressed (DE) genes between two successive developmental stages (>4-fold; 5% FDR). Note that most of the DE genes were unique in each comparison. (C) Functional enrichment analyses for DE genes between two successive developmental stages. Fifty-three GO terms—which can be broadly categorized into 7 biological processes—were assessed for degree of functional enrichment and were projected to two-dimensional semantic spaces. Only GO terms with a P value of < 0.05 are depicted in each panel.

and identified potentially novel transcripts that were absent in the reference annotations (Table 1). For identification of noncoding transcripts, the coding potential of the novel transcripts was computed by using the CPAT program (52). To maximize both sensitivity and specificity for noncoding transcript detection, the program was trained on the F. graminearum genome data set, and the threshold was set to a CPAT score of 0.540 (see Fig. D in Text S1) (cf. 0.364 for humans and 0.440 for mice [53]). The transcripts with a low coding potential (CPAT score of ≤0.540) were further scanned against the Pfam and Rfam databases to filter out transcripts encoding protein domains and harboring any known structural RNA motifs, respectively (E value of $<10^{-10}$ [Table 1]). Finally we only retained transcripts that were differentially expressed in at

TABLE 1 Identification of IncRNAs expressed during sexual development

Transcript class (code) ^a	No. of transcripts		
	Novel ^b	Noncoding ^c	DE ^d
Splicing variants (J)	5,185	92	17
Tandem transcripts (P)	740	388	100
Intergenic transcripts (U)	2,215	1,054	167
Antisense transcripts (X)	2,892	1,040	263
Sum	11,032	2,574	547

^aTranscript class codes were tagged by the *gffcompare* program (122).

^bAmong a total of 11 transcript class codes, transcripts tagged with the class codes J, P, U, and X were considered novel transcripts (see Text S1 for details).

^cNoncoding transcripts were identified by the coding potential assessment tool (CPAT) program (52) and further filtered by Pfam and Rfam database searches.

^dDifferentially expressed (DE) noncoding transcripts of at least one developmental stage were identified as F. graminearum IncRNAs.

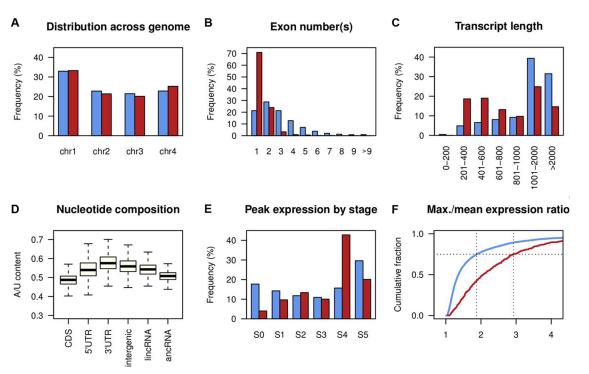


FIG 2 Genomic features of F. graminearum IncRNA. (A) Distribution of mRNA (blue bars) and IncRNA (red bars) across the four chromosomes. (B) Distribution of exon numbers per transcript. (C) Transcript length distribution. (D) A/U content of mRNA coding regions (CDS), 5' untranslated regions (UTRs), 3' UTRs, intergenic regions, long intergenic IncRNA (lincRNA), and antisense IncRNA (ancRNA). Box and whisker plots indicate the median, interquartile range between the 25th and 75th percentiles (box), and 1.5 interguartile range (whisker). (E) Distribution of developmental stages at which mRNA and IncRNA showed the highest expression level. (F) Cumulative distributions of ratios of maximum and mean expression values across the developmental stages. Blue lines indicate mRNA, and red lines indicate IncRNA.

least one developmental stage (5% FDR), yielding a total of 547 IncRNA candidates (Table 1; see Table S1 in the supplemental material).

The identified IncRNAs were distributed across the 4 chromosomes and were generally shorter, with fewer exons than mRNAs (transcripts with a CPAT score of >0.540) (Fig. 2A to C). Based on the relative position to mRNAs, IncRNAs can be classified as antisense IncRNA (ancRNA) or long intergenic ncRNA (lincRNA). There were 280 ancRNAs that overlapped more than 100 bp of an mRNA on the opposite strand and 237 lincRNAs that were situated between annotated genes (Table S1). The mean A/U content of ancRNA and lincRNA sequences falls between the coding sequences of the mRNAs and the intergenic regions (Fig. 2D). These distinctive genomic features of the lncRNA sequences are commonly observed in other eukaryotes (29, 54-57). However, we identified only 5 IncRNAs that showed similarity to IncRNAs in other eukaryotes (E value of $<10^{-10}$; see Table B in Text S1), suggesting either the poor status of IncRNA annotations in filamentous fungi or a high degree of sequence divergence in fungal IncRNAs.

Developmental expression of IncRNA. The sexual stage transcriptome data showed predominance of IncRNAs at the meiotic stage (S4), where the expression of many IncRNAs peaked (234 out of 547 [Fig. 2E]). We compared the degree of differential expression of IncRNAs to that of mRNAs (9,457 transcripts differentially expressed in at least one developmental stage at 5% FDR). The ratio of the maximum expression among six developmental stages to the mean expression over the remaining five stages was calculated for IncRNAs and the differentially expressed mRNAs. By this metric, IncRNA was prone to be more differentially expressed than mRNA ($P = 2.2 \times 10^{-16}$; Komolgorov-Smirnov test statistic, D = 0.39), with the third quartile value of the ratio measuring 2.93 for IncRNA and 1.88 for mRNA (Fig. 2F). Also, we identified seven coexpressed clusters of IncRNAs that showed developmental stage-specific expression

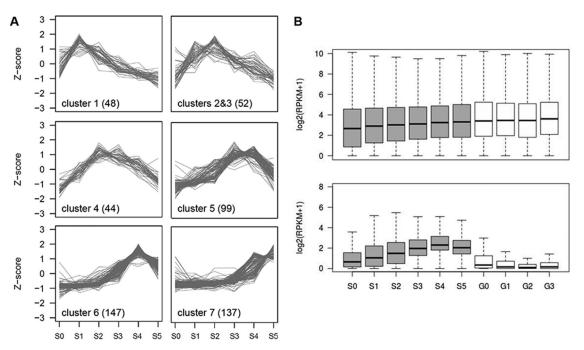


FIG 3 Sexual stage-induced IncRNA in F. graminearum. (A) Coexpressed clusters of IncRNAs. Trend plots of Z-score normalized expression values for IncRNAs (numbers in parentheses) in a given cluster are presented. (B) Expression distribution of mRNA (upper panel) and IncRNA (lower panel) for the sexual development transcriptome (gray boxes) and the vegetative growth transcriptome (white boxes). Box and whisker plots indicate the median, interquartile range between the 25th and 75th percentiles (box), and 1.5 interquartile range (whisker).

patterns, suggesting distinct roles of IncRNAs in different stages of perithecial development (Fig. 3A).

In addition to the sexual stage data set, we obtained transcriptome data during spore germination to investigate the degree of IncRNA expression in vegetative stages. The data set was comprised of four spore germination stages: G0, fresh spore; G1, polar growth; G2, doubling of long axis; and G3, branching of hyphae (see Fig. E in Text S1). Overall expression of the IncRNA gradually increased over the course of perithecial development, peaking at S4, while most of the IncRNA remained unexpressed or had low expression in the germination stages, indicating that most of the IncRNA expression is sexual stage specific (Fig. 3B).

Verification of IncRNA production. To validate IncRNA expression, we chose eight IncRNAs and performed PCR with 3' rapid amplification of cDNA ends (RACE-PCR) and Sanger sequencing. All the selected IncRNAs were amplified from total RNA extracts, and their polyadenylation sites were determined (see Fig. F in Text S1). Also, to examine if there is an intraspecific conservation in the IncRNA content and expression, we utilized degradome-seq data from another F. graminearum wild-type (WT) strain sampled at the meiotic stage (47). Degradation of IncRNA transcripts expressed at S4 was evident, which in turn confirmed the consistent IncRNA production in the two strains (Fig. 4; also see Fig. 5B below).

Fungal genomes are known for having shorter intergenic spaces than other eukaryotic genomes. Therefore, fungal IncRNAs could be a transcriptional noise arising from neighboring genes. To test this, we examined global patterns of the expression correlation between IncRNA and neighboring genes, with close examination of some selected examples whose expression was confirmed by 3' RACE-PCR. The lncRNAs antisense to ORC1, ORC2, and CENP-T each had an upstream gene in close proximity on the same strand (Fig. 4; see Fig. F in Text S1). However, the sexual stage expression between the lncRNAs and their respective upstream genes was not correlated (|r| < 0.50 [see Table C in Text S1]). Positive correlation was observed in expression levels between IncRNAs and their divergently transcribed genes as in the HIR1 and NSE4 loci

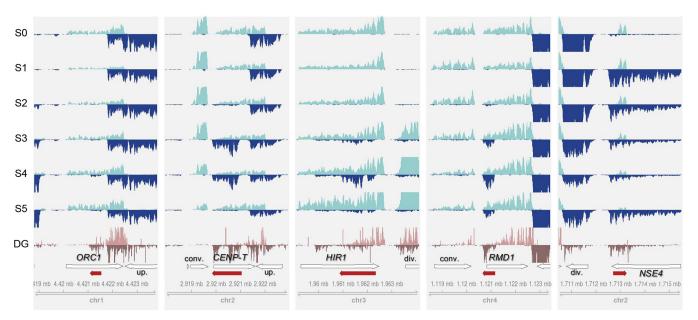


FIG 4 Examples of IncRNA expression across the sexual development stages. Per-base coverage of transcripts was plotted for both DNA strands in a 5-kb window. For the perithecial transcriptome data sets (S0 to S5), mapped reads of 3 biological replicates were pooled and then subsampled to 60 million reads for visual comparison of expression levels across the stages. For the degradome-seq data sets (DG), mapped reads of 2 replicates were pooled and displayed. The positions of IncRNAs (red arrows) and their neighboring genes (white arrows) are shown in the annotation track with the genome coordinate at the bottom of each panel. The genes overlapping IncRNAs on the opposite strand are labeled with abbreviated gene names in boldface: CENP-T, centromere protein T (FGRRES_16954), HIR1, histone regulatory protein 1 (FGRRES_05344); NSE4, nonstructural maintenance of chromosome element 4 (FGRRES_17018); ORC1, origin recognition complex subunit 1 (FGRRES_01336); RMD1, required for meiotic division 1 (FGRRES_06759). In relation to IncRNA position, neighboring genes are also labeled as follows: div., divergently transcribed gene on the opposite strand; conv., convergently transcribed gene on the opposite strand; up., upstream gene in tandem on the same strand.

(r > 0.8 [Fig. 4; see Table C in Text S1]), indicating prevalence of bidirectional promoters for IncRNA transcription (2, 58, 59). On the other hand, the expression of IncRNAs and their convergently transcribed genes tend not to be correlated, as in the CENP-T and RMD1 loci (|r| < 0.50 [Fig. 4; see Table C in Text S1]). These patterns were globally observed in IncRNA-associated loci (see Fig. F in Text S1), suggesting that the IncRNAs were not likely to be misannotated extensions of neighboring genes (54, 60, 61).

Identification of sRNA-enriched loci associated with IncRNAs. We found 300 sense mRNA-ancRNA pairs with different orientations: $5' \rightarrow 5'$ partial, $3' \rightarrow 3'$ partial, and full overlaps. One of the most common mechanisms involving antisense transcription is the RNA interference (RNAi) pathway incorporating small RNAs (sRNAs) generated from the double-stranded RNA regions. To investigate the degree and effect of sRNA production in the ancRNA loci, we analyzed the previously published sRNA-seq and degradome-seq data at the meiotic stage (47). As expected, sRNA reads were mapped at a higher frequency to ancRNAs than to mRNAs without overlapping antisense transcripts, the sense mRNAs, or lincRNAs ($P < 1.2 \times 10^{-8}$; Kruskal-Wallis test statistic, H = 155 [Fig. 5A]), suggesting that the ancRNA loci may serve as a major source for endogenous sRNA production. However, the correlation of the degradome-seg and our RNA-seq data at S4 showed that sRNA-mediated endonucleolytic cleavage of ancRNAs and sense mRNAs was comparable to that of mRNAs without antisense transcripts (Fig. 5B), implying that the ancRNA loci were not preferentially targeted by RNAi machinery, posttranscriptionally.

It remains paradoxical that gene silencing induced by heterochromatin formation requires sRNA production via cotranscriptional processes, sometimes from lncRNAs (62-66). To search for any IncRNAs associated with sRNA-enriched loci that could be indicative of transcriptional gene silencing events, we examined the top 80 sRNA clusters ranked by the number of mapped reads, which accounted for 62% of mapped sRNA reads (see Table D in Text S1). Production of sRNAs in the top 80 clusters was dependent on Dicers and Argonautes (47), indicating that the sRNAs were produced by

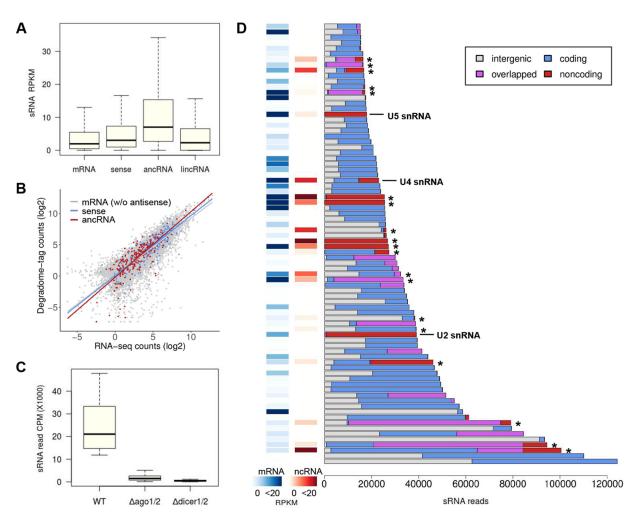


FIG 5 IncRNA associated with small RNA-enriched loci. (A) sRNA reads mapped to mRNAs without antisense transcripts (10,928 loci), sense mRNAs for ancRNAs (295 loci), ancRNAs (276 loci), and lincRNA (235 loci) are represented as RPKM. The number of sRNA reads aligned to ancRNA loci was more than those of the other classes of transcripts (Benjamini-Hochberg adjusted P value of <0.0001; Dunn's pairwise multiple comparisons). (B) Correlation analysis between transcript abundance and degradome tag count at the meiotic stage. Lines depict regressions for different classes of transcripts. (C) sRNA reads mapped to the top 80 sRNA clusters in different genotypes are represented as counts per million (CPM). The numbers of sRNA reads mapped to the clusters were significantly reduced in the \(\Delta dicer1/2 \) mutant, with double deletion of Dicer genes (FGRRES_09025 and FGRRES_04408), and the $\Delta ago 1/2$ mutant, with double deletion of Argonaute genes (FGRRES_16976 and FGRRES_00348) (47). (D) Fractions of sRNA reads mapped to intergenic regions (white), coding genes (blue), overlapped regions (purple), and noncoding genes (red), including IncRNAs marked with asterisks in the top 80 sRNA clusters. Overlapped regions were defined if coding or noncoding genes in the region were present on the both sides of DNA in the de novo annotations. Expression values (RPKM) for the closest coding gene (mRNA) to the center of each sRNA cluster are shown as heat maps, along with expression values for ncRNAs, if any, that were present in the same cluster.

RNAi machinery (Fig. 5C). Most of the sRNA clusters were found in genic regions, containing at least one annotated gene, to which a large portion of sRNA reads were mapped (Fig. 5D). We observed that the coding genes closest to the centers of sRNA clusters exhibited overall low expression (<0.5 read per kilobase per million [RPKM] in 22 out of the 80 clusters) (Fig. 5D). A significant portion of sRNAs were also derived from noncoding transcripts and genic regions overlapped with antisense transcripts, some of which were identified as IncRNAs (Fig. 5D; see Fig. S1 in the supplemental material). Unexpectedly, the IncRNAs associated with sRNA clusters exhibited moderate expression (n = 19; median expression of 5.2 RPKM). In addition, the coding genes closest to the centers of the sRNA clusters showed higher expression levels (n = 19; median expression of 4.0 RPKM) than those without an associated lncRNA (n = 61; median expression of 1.3 RPKM; P = 0.029; Mann-Whitney test statistic, U = 747).

Coexpression of IncRNAs and their sense transcripts. We could not find strong evidence for sRNA-mediated transcriptional and posttranscriptional gene silencing in

mBio[®]

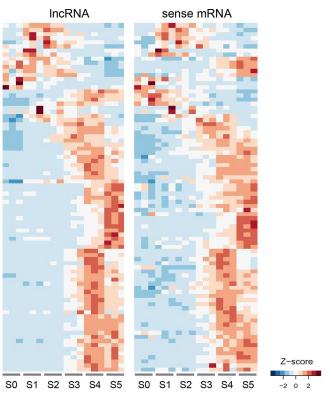


FIG 6 Parallel induction of sense mRNA and antisense IncRNA pairs during sexual development. Expression data of sense mRNA and antisense IncRNA pairs in 18 samples for the perithecia transcriptome (S0 to S5) were reordered by the BLIND program. The RPKM values of sense mRNA and antisense IncRNA pairs with absolute Pearson's correlation greater than 0.7 (P < 0.05; Fisher's exact test) were clustered by Euclidean distance, and heat maps of Z-score normalized RPKM values are presented.

the ancRNA loci. Interestingly, we did observe gene expression correlation in many ancRNAs and sense mRNA pairs across the sexual stages (85 out of the 300 pairs with Pearson's correlation |r| > 0.70; Fisher's exact test, P < 0.05 [Fig. 6]), most of which were positively correlated (76 out of the 85 pairs). We asked whether the ancRNAs are antisense to genes involved in a specific biological process. We found that 17 genes in the ancRNA loci were involved in DNA metabolic processes such as DNA replication, repair, and recombination (Table 2). Notably, the positively correlated sense mRNAs were enriched for the GO term "DNA metabolism" at a 5% FDR (8 out of the 76 genes [see Table E in Text S1]), and among them, five genes (FGRRES_00736, FGRRES_01614, FGRRES 06286, FGRRES 07280, and FGRRES 08658) appeared to have functions related to DNA damage repair (Table 2). In addition, we surveyed previously published research for the 295 sense genes to determine if there are any genes known to affect sexual development. We found one gene, RGSC (FGRRES_09585; encoding a regulator of G proteins), whose knockout phenotype was defective in ascospore morphology (67).

Identification of IncRNAs affected by the NMD pathway. The NMD pathway is known to regulate the expression of noncoding genes, as well as coding genes, for sexual sporulation in yeasts. To identify the sexual stage-induced IncRNAs that were affected by NMD, we generated a deletion strain lacking XRN1 (FGRRES_06799) that encodes a 5' \rightarrow 3' exonuclease. The $\Delta xrn1$ deletion mutant displayed a lower growth rate on different growth media compared to the wild type (WT); however, it produced normal-shaped asexual spores (Fig. 7A; see Fig. G in Text S1). We observed that perithecial development was significantly delayed and ascus development was defective in the $\Delta xrn1$ strain, producing variable ascospore sizes (Fig. 7A; see Fig. G in Text S1). The defective sexual phenotype was restored by genetic complementation with XRN1 (see Fig. G in Text S1).

TABLE 2 Sense transcripts encoding DNA metabolism-related gene products

Gene ID no.	Putative function	Protein domain (E value) ^a	Correlation with ancRNAb
FGRRES_00667	Similar to chromatin assembly complex subunit p60	Pfam12894 (3.69e-08)	0.76 (IncRNA-026)
FGRRES_00736	Similar to DNA damage response protein WSS1	Pfam08325 (3.75e-79)	0.89 (IncRNA-030)
FGRRES_01336	Origin recognition complex subunit 1	Cd04720 (3.48e-37)	0.69 (IncRNA-065)
FGRRES_01614	Similar to DNA mismatch repair protein MLH3	COG0323 (5.15e-39)	0.88 (IncRNA-083)
FGRRES_05344	Histone regulatory protein 1 (HIR1)	Pfam07569 (6.27e-99)	0.29 (IncRNA-324)
FGRRES_05737	ssDNA/RNA-binding protein	Pfam00076 (1.73e-15)	0.91 (lncRNA-342)
FGRRES_05781	Similar to TCB2 transposase	Pfam01498 (3.32e-08)	0.97 (IncRNA-344)
FGRRES_06122	Origin recognition complex subunit 2	Pfam04084 (3.98e-163)	0.39 (IncRNA-356)
FGRRES_06231	Origin recognition complex subunit 4	Pfam14629 (1.82e-70)	0.01 (IncRNA-367)
FGRRES_06286	Similar to DNA repair polymerase RAD30	Cd01703 (6.27e-145)	0.75 (IncRNA-369)
FGRRES_06339	DNA primase small subunit	Pfam01896 (8.59e-42)	0.40 (IncRNA-373)
FGRRES_07280	Similar to DNA repair helicase ERCC3/RAD25	Pfam16203 (6.04e-173)	0.70 (IncRNA-461)
FGRRES_08650	Similar to SEN1 helicase	Pfam12726 (0e+00)	0.66 (IncRNA-214)
FGRRES_08658	Similar to DNA repair endonuclease ERCC4/RAD16	Pfam02732 (2.80e-25)	0.77 (IncRNA-216)
FGRRES_09872	Similar to DNA repair dioxygenase AlkB	COG3145 (4.37e-07)	-0.02 (IncRNA-144)
FGRRES_16954	Centromere protein T (CENP-T)	Pfam15511 (6.41e-47)	-0.36 (IncRNA-235)
FGRRES_17018	Nonstructural maintenance of chromosome element 4 (NSE4)	Pfam08743 (3.83e-45)	0.10 (IncRNA-201)

^aBLAST search for associated protein domains related to DNA metabolic processes.

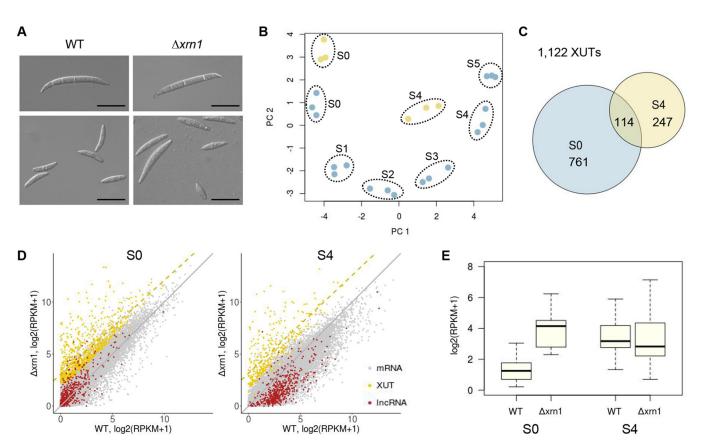


FIG 7 Identification of IncRNAs regulated by the NMD pathway. (A) Asexual spore (macroconidia) morphology (upper panels) and sexual spore (ascospore) morphology (lower panels). Note the size variation of ascospores in the $\Delta xm1$ mutant. Scale bars = 20 μ m. (B) Principal-component analysis of the perithecial transcriptome data from the wild type (S0 to S5 [blue circles]) and $\Delta xrn1$ transcriptome data (S0 and S4 [yellow circles]). (C) Venn diagram showing the overlap of Xrn1-sensitive unstable transcripts (XUTs) at S0 and S4 (D) Two-dimensional plots of different classes of transcripts between the wild-type (abscissa) and $\Delta xrn1$ mutant (ordinate) expression data at S0 and S4 (diagonal [gray line]). The yellow dashed line shows regression for the XUTs. (E) Expression distribution of 25 IncRNAs that were identified as XUTs at S0 and S4. Box and whisker plots indicate the median, interquartile range between the 25th and 75th percentiles (box), and 1.5 interquartile range (whisker).

Pearson's correlation for the expression levels of sense transcripts and ancRNAs across the sexual development stages. Values greater than 0.7 are marked in boldface.

mBio

To identify IncRNA whose expression was affected by NMD activity, we analyzed the transcriptome data from the WT and the Δxrn1 strain at the vegetative (S0) and meiotic (S4) stages. The transcriptome data of the $\Delta xrn1$ strain were distinct from, but most similar to, those of WT at the corresponding stages, indicating drastic effects of XRN1 on gene expression levels as a major component for RNA turnover (Fig. 7B). After expression values were normalized with 124 ribosomal protein genes that were known to be relatively insensitive to NMD activity (26), we identified a total of 1,122 XUTs that were differentially expressed at S0 or S4 in the $\Delta xrn1$ strain (>3-fold increase at 5% FDR [Fig. 7C; see Table S2 in the supplemental material]). After the normalization, the XUTs identified at S0 and S4 showed median 5.0- and 5.8-fold increases in the $\Delta xrn1$ mutant, respectively (Fig. 7D; Table S2). Many XUTs were previously annotated transcripts or isoforms of them (70% [781/1,122]), and only 11% of XUTs (122/1,122) were predicted to be noncoding transcripts (CPAT score of ≤0.540) (Table S2). Among the noncoding XUTs, we identified 25 IncRNAs whose expression was elevated upon the Xrn1 depletion at SO and showed increasing patterns across the sexual stages (Fig. 7E; see Fig. S2 in the supplemental material). In addition, many of the coding XUTs identified at S0 were also induced as the sexual development progressed (see Fig. H in Text S1). Interestingly, we found that key components of the RNA-induced silencing complex (RISC) (68) were also identified as XUTs, such as DICER2 (FGRRES_04408), AGO1 (FGRRES_16976), Neurospora crassa QIP homologue (FGRRES_06722), and RDRP genes (coding for RNA-dependent RNA polymerases [FGRRES_01582, _04619, and _09076]).

DISCUSSION

Here we profiled transcriptomes of vegetative and sexual stages that span the entire life cycle of F. graminearum. IncRNAs are usually an order of magnitude less abundant than mRNAs (69, 70), so the unprecedented depth of the sequencing data we generated (total of 938 million mapped reads) enabled us to capture scantly expressed noncoding transcripts. This study has revealed global properties of IncRNAs during perithecial development, characterized by dynamic and developmental stage-specific expression. In the last step of our IncRNA identification pipeline, we only included differentially expressed noncoding transcripts to discern IncRNAs of biological significance. This filtering step allowed us to identify low-abundance IncRNAs that alone could be argued to be transcriptional noise (71, 72). For many antisense IncRNAs, we detected parallel induction with their respective sense transcripts across sexual development and identified a subset of IncRNAs that are sensitive to NMD activity before sexual induction.

Our lists of F. graminearum IncRNAs contain many confident IncRNA annotations, but are still far from complete. This is primarily due to difficulty in unequivocal determination of whether a transcript is coding or noncoding (73). Although protein products from most of the bona fide IncRNAs with predicted ORFs have not been detected in cells (74, 75), many IncRNAs with one or more ORFs have been shown to be associated with ribosomes in budding yeast, Arabidopsis, and animals (76-81). By scrutinizing the sRNA-enriched loci, we identified several transcripts with short ORFs (ranging from 49 to 227 amino acids) as putative IncRNAs. These novel transcripts were initially filtered out by the CPAT program; however, another annotation tool, CPC2 (82), classified them as IncRNAs, and the IncRNA annotations were also supported by the lack of cross-species conservation of the deduced polypeptides (E value of $\geq 10^{-10}$ [see Table F in Text S11).

Another source of false negatives in our data set may have arisen from poly(A)based library preparation that excluded some lncRNAs lacking poly(A) tail. This may account for the inclusion of fewer IncRNAs in our XUT identification process, as deadenylation often takes place prior to $5' \rightarrow 3'$ degradation by Xrn1. More precise identification of IncRNAs, especially for those undergoing deadenylation by the RNA quality control machineries, should be accompanied by cap analysis gene expression (CAGE)-seq (83–85). However, most lncRNAs identified in this study were also detected

by degradome-seq (47), another sequencing method (also known as parallel analysis of RNA ends [86]), which further validates their authenticity.

To better understand how the expression of IncRNAs was controlled during perithecial development, we identified XUTs, as the NMD pathway has been shown to determine the fate of IncRNAs (21, 84). However, only 5% of the sexual stage-induced IncRNAs (25/547) were affected by Xrn1 activity, and the majority of the IncRNAs seemed to escape from the RNA surveillance system. The proportion of XUTs among the IncRNAs in F. graminearum is comparable to that in other eukaryotes, where approximately 4 to 14% of noncoding transcripts are known to undergo NMD (25). Interestingly, sexually induced RISC components were highly upregulated in the Δxrn1 mutant, indicating a possible link between the gene silencing pathway and the NMD pathway. Further studies on IncRNAs and the RNAi machinery that were affected by NMD activity will provide novel insights into regulatory mechanisms via altered RNA metabolism during sexual development.

We initially aimed to dissect the roles of other exosome and NMD components in IncRNA expression, such as RRP6 (FGRRES_06049), DBP2 (FGRRES_16145), and MTR4 (FGRRES_01656) (84, 87). However, efforts to obtain deletion strains lacking either the DBP2, MTR4, or RRP6 gene were unsuccessful (W. Kim and F. Trail, unpublished data), suggesting these genes may be essential in F. graminearum, as they are in another filamentous fungus, Neurospora crassa (88, 89). In fission yeast, the nuclear exosome mediates gene silencing of coding and noncoding loci during vegetative growth, and the sexual sporulation is triggered by disassembly of heterochromatin on the silenced loci (90). Therefore, it will be interesting to see if the regulation of IncRNA expression is achieved by modulation of chromatin status before and after the sexual induction in F. graminearum.

Antisense transcription can influence synthesis, expression kinetics, and stability of sense transcripts through a variety of mechanisms (5, 91). Antagonism of IncRNAs against meiotic gene expression has been documented in yeasts (10, 11, 92-95). Nevertheless, there has been growing evidence that IncRNAs activate repressed genes by modulating local chromatin structures, which can facilitate coordinated gene expression in budding yeast (96, 97) and other eukaryotes (98-101). In favor of this regulatory phenomenon, we hypothesize that the F. graminearum IncRNAs, which showed parallel induction with DNA metabolism-related genes across the sexual development stages, may play a role during meiosis that requires DNA break and repair for homologous recombination, according to "quilt by association" (1, 29). However, IncRNAs often exhibited cell-type-specific expression (100, 102, 103) and even allelespecific expression within the same nucleus (104). Therefore, we cannot rule out the possibility that the positively correlated pairs of IncRNA and sense transcript are in fact mutually exclusively expressed in different tissues or cell types in the perithecia, inhibiting each other.

Although genes found in sRNA-enriched loci are usually silenced by RNAi-dependent heterochromatin formation, some of the sRNA clusters involving IncRNAs, paradoxically, were transcriptionally active at the meiotic stage in F. graminearum. In our developmental transcriptome data, we observed that many stage-specific genes were induced over the course of sexual development, which may play key roles in the corresponding perithecial tissue development. It is conceivable that the silenced genes in vegetative tissues were derepressed by IncRNA expression by yet unknown mechanisms upon sexual induction. In N. crassa, molecular studies on the circadian clock gene locus revealed a periodic gene activation and repression mechanism involving IncRNA and facultative heterochromatin formation. The expression of an IncRNA antisense to the circadian clock gene promotes sense transcription by generating a more transcriptionally permissive chromatin that has been silenced by sRNA-mediated heterochromatin formation (105).

This study presents genome-wide characterization of IncRNAs during fruiting body development. The transcriptome landscape, including IncRNAs during the life cycle of F. graminearum, provides fundamental genomic resources to the Fusarium community. The detailed molecular study of newly identified lncRNAs, with its established tools for

rapid genetic analyses and ample genetic resources, will contribute to the understanding of how fungi utilize noncoding genomes for laying out their multicellular body plan.

MATERIALS AND METHODS

Data generation and processing. The *F. graminearum* genome assembly (39) and the Ensembl annotation version 32 (40) of wild-type strain PH-1 (accession no. FGSC 9075 and NRRL 31084) were used throughout this study. For total RNA extraction, synchronized fungal tissues were collected from carrot agar cultures at the previously defined developmental stages during perithecial formation (36–38, 48). For transcriptome data of spore germination stages, asexual spores (macroconidia) were spread on Bird agar medium (106) overlaid with a cellophane membrane and sampled at the indicated spore germination stages (see Fig. E in Text S1). Strand-specific cDNA libraries were constructed from poly(A)-captured RNAs, using the Kapa stranded RNA-Seq library preparation kit (Kapa Biosystems, Wilmington, MA), and sequenced on the Illumina HiSeq 2500 platform (Illumina, Inc., San Diego, CA) at the Michigan State University Research Technology Support Facility (https://rtsf.natsci.msu.edu/genomics). Following quality control for raw reads (Text S1), filtered reads were mapped to the repeat-masked genome using the HISAT2 program (v2.0.4) (107), and a genome-guided transcriptome assembly was performed using the StringTie program (v1.3.0) to generate *de novo* transcript annotations (108).

Differential expression and functional enrichment analyses. Read counts for gene loci were calculated using the *htseq-count* program (v0.6.1) (109). On average, 87% of mapped reads were overlapped to exons in the *de novo* annotations. Gene expression levels in counts per million (CPM) values were computed and normalized by effective library size estimated by trimmed mean of *M* values, using the edgeR R package (v3.14.0) (110). Only genes with CPM values greater than 1 in at least 3 samples were kept for further analyses (15,476 out of 20,459 gene loci). Then, differentially expressed (DE) genes showing greater than 4-fold difference at an FDR of 5% were identified between two successive developmental stages, using the limma R package (v3.28.21) (111). GO terms were assigned to the *de novo* annotations, using the Trinotate program (v3.0.1) (112). The list of GO terms was customized by adding several GO terms related to developmental processes to the GO Slim terms specific for fission yeast (113). Functional enrichment analyses for DE genes were performed using the GOseq R package (v1.24.0), including only those genes annotated by one or more GO terms (114). To assess enrichment of GO terms, the Wallenius approximation (an extension of the hypergeometric distribution) and Benjamini-Hochberg method were used to calculate the FDR-corrected *P* value.

Conserved IncRNAs search. To search for conserved IncRNAs, the 547 IncRNAs were queried against the RNAcentral database version 5 (http://rnacentral.org), using the *nhmmer* program, which detects remote homologies, in the HMMER software (v3.1b2) (115). Search hits with an E value of $<10^{-10}$ were reported.

Coexpression network analysis. The weighted gene correlation network analysis (WGCNA) R package (v1.51) (116) was used to cluster IncRNAs by averaged RPKM values for developmental stages. The "pickSoftThreshold" function was used to determine soft-thresholding power that measures the strength of correlation based on not just the direct correlation value of pairs of genes, but also the weighted correlations of all of their shared neighbors in the network space (117). The soft-thresholding power of 26 was selected, which is the lowest power for which the scale-free topology model fit index reaches 0.80. A range of treecut values were tested for cluster detection, and the value was set to 0.18 (corresponding to a correlation of 0.82). All other WGCNA parameters remained at their default settings.

Identification of small RNA clusters. The sRNA-seq and degradome-seq data were obtained from NCBI GEO (GSE87835) and NCBI SRA (PRJNA348145), respectively (47). In filamentous fungi, the size of a majority of sRNAs ranges from 17 to 27 nucleotides (nt), with a strong 5' U preference (70 to 82%) (47, 118). Thus, clusters of 17- to 27-nt-long 5'-U sRNA reads were detected across the genome, using the ShortStack program (v3.8.2) (119) with the option arguments "--pad 22" and "--mincov 20." Subsequently, the number of 5' U sRNA reads aligned to different genomic features (e.g., coding regions) were counted for each sRNA cluster, using the *htseq-count* program (v0.8.0) (109). The degradome-seq data set was processed according to the previous study (47).

Expression correlation analysis. The expression value matrix for the 18 RNA-seq data (6 stages \times 3 replicates) were rearranged by the BLIND program (see Fig. B in Text S1). Pearson's correlation and the associated P value by Fisher's exact test were calculated for the expression levels of the 300 sense mRNA-antisense lncRNA pairs, using an R script (29). For sense mRNAs only, which showed positive correlation with antisense lncRNAs, we performed functional enrichment analyses, using the GOseq R package (v1.24.0) (114), as done for DE genes between two successive developmental stages (see above).

XUT identification. Generation, confirmation, and complementation of the $\Delta xrn1$ mutant are described in Text S1, and the primers used in this study are listed in Table S3 in the supplemental material. To incorporate novel transcripts that were only expressed in the $\Delta xrn1$ mutant due to loss of NMD activity, the RNA-seq reads for the $\Delta xrn1$ mutant were independently processed, and the assembled transcripts were merged to the *de novo* annotations, using the "merge" function in the StringTie program (v1.3.0) (108). Transcript abundance is globally affected by XRN1 deletion in budding yeast; however, the expression of ribosomal protein genes shows only slight increases in the $\Delta xrn1$ mutant and remains constant after lithium treatment, which inhibits $5' \rightarrow 3'$ exonuclease activities (26, 120). Thus, before XUT identification, expression values were normalized in such a way that ribosomal protein genes are expressed at the same levels in the WT and $\Delta xrn1$ mutant by multiplying by 0.66 and 1.96 the RPKM values of the WT at S0 and S4, respectively (see Fig. I in Text S1). Differential expression analyses were performed, using the "stattest" function with the option argument "libadjust = FALSE" in the Ballgown R package (v2.8.4) (121). The DE transcripts showing a 3-fold increase in the $\Delta xrn1$ mutant were identified

Kim et al.

as XUTs (5% FDR). We excluded transcripts with relatively low expression levels in the $\Delta xrn1$ mutant (<3.0 RPKM) from the putative XUTs to reduce possible false positives.

Accession number(s). The RNA-seq data generated in the present work have been deposited in NCBI's Gene Expression Omnibus (https://www.ncbi.nlm.nih.gov/geo) and are accessible through GEO series accession no. GSE109095, which is composed of the two data sets—the sexual stage and $\Delta xrn1$ mutant data set (GSE109094) and the spore germination stage data set (GSE109088).

SUPPLEMENTAL MATERIAL

Supplemental material for this article may be found at https://doi.org/10.1128/mBio .01292-18.

TEXT S1, PDF file, 2.6 MB.

FIG S1, PDF file, 0.7 MB.

FIG S2, PDF file, 1.2 MB.

TABLE S1, RTF file, 0.1 MB.

TABLE S2, RTF file, 0.1 MB.

TABLE S3, DOCX file, 0.1 MB.

ACKNOWLEDGMENTS

We thank Brad Cavinder, Kevin Childs, and Nicholas Panchy (Michigan State University, East Lansing, MI) for helpful discussion on analyzing data and for providing Perl and Python scripts.

This publication is based upon work supported by the National Science Foundation under grant no. 1456482 to F.T. and Michigan AgBioResearch (to F.T.). Also, this work was supported by Program Area: Plant Health and Production and Plant Products under grant no. 2015-67013-22932 to F.T. from the USDA National Institute of Food and Agriculture. The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

Any opinions, findings, conclusions, or recommendations expressed in this publication are those of the authors and do not necessarily reflect the view of the U.S. Department of Agriculture and the National Science Foundation.

REFERENCES

- Guttman M, Amit I, Garber M, French C, Lin MF, Feldser D, Huarte M, Zuk O, Carey BW, Cassady JP, Cabili MN, Jaenisch R, Mikkelsen TS, Jacks T, Hacohen N, Bernstein BE, Kellis M, Regev A, Rinn JL, Lander ES. 2009. Chromatin signature reveals over a thousand highly conserved large non-coding RNAs in mammals. Nature 458:223–227. https://doi.org/10 .1038/nature07672.
- 2. Xu Z, Wei W, Gagneur J, Perocchi F, Clauder-Münster S, Camblong J, Guffanti E, Stutz F, Huber W, Steinmetz LM. 2009. Bidirectional promoters generate pervasive transcription in yeast. Nature 457:1033–1037. https://doi.org/10.1038/nature07728.
- Kapranov P, Cheng J, Dike S, Nix DA, Duttagupta R, Willingham AT, Stadler PF, Hertel J, Hackermüller J, Hofacker IL, Bell I, Cheung E, Drenkow J, Dumais E, Patel S, Helt G, Ganesh M, Ghosh S, Piccolboni A, Sementchenko V, Tammana H, Gingeras TR. 2007. RNA maps reveal new RNA classes and a possible function for pervasive transcription. Science 316:1484–1488. https://doi.org/10.1126/science .1138341.
- Rinn JL, Chang HY. 2012. Genome regulation by long noncoding RNAs. Annu Rev Biochem 81:145–166. https://doi.org/10.1146/annurev-biochem -051410-092902
- Quinn JJ, Chang HY. 2016. Unique features of long non-coding RNA biogenesis and function. Nat Rev Genet 17:47–62. https://doi.org/10 .1038/nrg.2015.10.
- Brockdorff N, Ashworth A, Kay GF, McCabe VM, Norris DP, Cooper PJ, Swift S, Rastan S. 1992. The product of the mouse *Xist* gene is a 15 kb inactive X-specific transcript containing no conserved ORF and located in the nucleus. Cell 71:515–526. https://doi.org/10.1016/0092-8674 (92)90519-1
- Fatica A, Bozzoni I. 2014. Long non-coding RNAs: new players in cell differentiation and development. Nat Rev Genet 15:7–21. https://doi. org/10.1038/nrg3606.
- 8. Flynn RA, Chang HY. 2014. Long noncoding RNAs in cell fate program-

- ming and reprogramming. Cell Stem Cell 14:752–761. https://doi.org/10.1016/j.stem.2014.05.014.
- Hiriart E, Verdel A. 2013. Long noncoding RNA-based chromatin control of germ cell differentiation: a yeast perspective. Chromosome Res 21:653–663. https://doi.org/10.1007/s10577-013-9393-5.
- van Werven FJ, Neuert G, Hendrick N, Lardenois A, Buratowski S, van Oudenaarden A, Primig M, Amon A. 2012. Transcription of two long non-coding RNAs mediates mating type control of gametogenesis in budding yeast. Cell 150:1170–1181. https://doi.org/10.1016/j.cell.2012 .06.049.
- Hongay CF, Grisafi PL, Galitski T, Fink GR. 2006. Antisense transcription controls cell fate in *Saccharomyces cerevisiae*. Cell 127:735–745. https://doi.org/10.1016/j.cell.2006.09.038.
- Gelfand B, Mead J, Bruning A, Apostolopoulos N, Tadigotla V, Nagaraj V, Sengupta AM, Vershon AK. 2011. Regulated antisense transcription controls expression of cell-type-specific genes in yeast. Mol Cell Biol 31:1701–1709. https://doi.org/10.1128/MCB.01071-10.
- Harigaya Y, Tanaka H, Yamanaka S, Tanaka K, Watanabe Y, Tsutsumi C, Chikashige Y, Hiraoka Y, Yamashita A, Yamamoto M. 2006. Selective elimination of messenger RNA prevents an incidence of untimely meiosis. Nature 442:45–50. https://doi.org/10.1038/nature04881.
- Hiriart E, Vavasseur A, Touat-Todeschini L, Yamashita A, Gilquin B, Lambert E, Perot J, Shichino Y, Nazaret N, Boyault C, Lachuer J, Perazza D, Yamamoto M, Verdel A. 2012. Mmi1 RNA surveillance machinery directs RNAi complex RITS to specific meiotic genes in fission yeast. EMBO J 31:2296–2308. https://doi.org/10.1038/emboj.2012.105.
- Yamashita A, Shichino Y, Tanaka H, Hiriart E, Touat-Todeschini L, Vavasseur A, Ding DQ, Hiraoka Y, Verdel A, Yamamoto M. 2012. Hexanucleotide motifs mediate recruitment of the RNA elimination machinery to silent meiotic genes. Open Biol 2:120014. https://doi.org/10.1098/rsob.120014.
- 16. Ding DQ, Okamasa K, Yamane M, Tsutsumi C, Haraguchi T, Yamamoto

- M, Hiraoka Y. 2012. Meiosis-specific noncoding RNA mediates robust pairing of homologous chromosomes in meiosis. Science 336:732-736. https://doi.org/10.1126/science.1219518.
- 17. Davis CA, Ares M. 2006. Accumulation of unstable promoter-associated transcripts upon loss of the nuclear exosome subunit Rrp6p in Saccharomyces cerevisiae. Proc Natl Acad Sci U S A 103:3262-3267. https://doi .org/10.1073/pnas.0507783103.
- 18. Camblong J, Iglesias N, Fickentscher C, Dieppois G, Stutz F. 2007. Antisense RNA stabilization induces transcriptional gene silencing via histone deacetylation in S. cerevisiae. Cell 131:706-717. https://doi.org/ 10.1016/i.cell.2007.09.014.
- 19. Lardenois A, Liu Y, Walther T, Chalmel F, Evrard B, Granovskaia M, Chu A. Davis RW, Steinmetz LM, Primig M. 2011. Execution of the meiotic noncoding RNA expression program and the onset of gametogenesis in yeast require the conserved exosome subunit Rrp6. Proc Natl Acad Sci U S A 108:1058-1063. https://doi.org/10.1073/pnas.1016459108.
- 20. Preker P, Nielsen J, Kammler S, Lykke-Andersen S, Christensen MS, Mapendano CK, Schierup MH, Jensen TH. 2008. RNA exosome depletion reveals transcription upstream of active human promoters. Science 322:1851-1854. https://doi.org/10.1126/science.1164096.
- 21. Smith JE, Baker KE. 2015. Nonsense-mediated RNA decay—a switch and dial for regulating gene expression. BioEssays 37:612-623. https:// doi.org/10.1002/bies.201500007.
- 22. Niazi F, Valadkhan S. 2012. Computational analysis of functional long noncoding RNAs reveals lack of peptide-coding capacity and parallels with 3' UTRs. RNA 18:825–843. https://doi.org/10.1261/rna.029520.111.
- 23. Kurihara Y, Matsui A, Hanada K, Kawashima M, Ishida J, Morosawa T, Tanaka M, Kaminuma E, Mochizuki Y, Matsushima A, Toyoda T, Shinozaki K, Seki M. 2009. Genome-wide suppression of aberrant mRNA-like noncoding RNAs by NMD in Arabidopsis. Proc Natl Acad Sci U S A 106:2453-2458. https://doi.org/10.1073/pnas.0808902106.
- 24. Tani H, Torimura M, Akimitsu N. 2013. The RNA degradation pathway regulates the function of GAS5 a non-coding RNA in mammalian cells. PLoS One 8:e55684. https://doi.org/10.1371/journal.pone.0055684.
- 25. Ruiz-Orera J, Messeguer X, Subirana JA, Alba MM. 2014. Long noncoding RNAs as a source of new peptides. eLife 3:e03523. https://doi .org/10.7554/eLife.03523.
- 26. van Dijk EL, Chen CL, d'Aubenton-Carafa Y, Gourvennec S, Kwapisz M, Roche V, Bertrand C, Silvain M, Legoix-Né P, Loeillet S, Nicolas A, Thermes C, Morillon A. 2011. XUTs are a class of Xrn1-sensitive antisense regulatory non-coding RNA in yeast. Nature 475:114-117. https://doi.org/10.1038/nature10118.
- 27. Mattick JS, Taft RJ, Faulkner GJ. 2010. A global view of genomic information-moving beyond the gene and the master regulator. Trends Genet 26:21-28. https://doi.org/10.1016/j.tig.2009.11.002.
- 28. Liu G, Mattick JS, Taft RJ. 2013. A meta-analysis of the genomic and transcriptomic composition of complex life. Cell Cycle 12:2061-2072. https://doi.org/10.4161/cc.25134.
- 29. Gaiti F, Fernandez-Valverde SL, Nakanishi N, Calcino AD, Yanai I, Tanurdzic M, Degnan BM. 2015. Dynamic and widespread IncRNA expression in a sponge and the origin of animal complexity. Mol Biol Evol 32:2367-2382. https://doi.org/10.1093/molbev/msv117.
- 30. Knoll AH. 2011. The multiple origins of complex multicellularity. Annu Rev Earth Planet Sci 39:217–239. https://doi.org/10.1146/annurev.earth .031208.100209
- 31. Nguyen TA, Cissé OH, Yun Wong JY, Zheng P, Hewitt D, Nowrousian M, Stajich JE, Jedd G. 2017. Innovation and constraint leading to complex multicellularity in the Ascomycota. Nat Commun 8:14444. https://doi .org/10.1038/ncomms14444.
- 32. Bumgarner SL, Dowell RD, Grisafi P, Gifford DK, Fink GR. 2009. Toggle involving cis-interfering noncoding RNAs controls variegated gene expression in yeast. Proc Natl Acad Sci U S A 106:18321-18326. https:// doi.org/10.1073/pnas.0909641106.
- 33. Chacko N, Zhao Y, Yang E, Wang L, Cai JJ, Lin X. 2015. The IncRNA RZE1 controls cryptococcal morphological transition. PLoS Genet 11: e1005692. https://doi.org/10.1371/journal.pgen.1005692.
- 34. Trail F. 2009. For blighted waves of grain: Fusarium graminearum in the postgenomics era. Plant Physiol 149:103-110. https://doi.org/10.1104/ pp.108.129684.
- 35. Ma LJ, Geiser DM, Proctor RH, Rooney AP, O'Donnell K, Trail F, Gardiner DM, Manners JM, Kazan K. 2013. Fusarium pathogenomics. Annu Rev Microbiol 67:399-416. https://doi.org/10.1146/annurev-micro-092412 -155650
- 36. Hallen HE, Huebner M, Shiu SH, Güldener U, Trail F. 2007. Gene expres-

- sion shifts during perithecium development in Gibberella zeae (anamorph Fusarium graminearum), with particular emphasis on ion transport proteins. Fungal Genet Biol 44:1146-1156. https://doi.org/10 .1016/j.fgb.2007.04.007.
- 37. Sikhakolli UR, López-Giráldez F, Li N, Common R, Townsend JP, Trail F. 2012. Transcriptome analyses during fruiting body formation in Fusarium graminearum and Fusarium verticillioides reflect species life history and ecology. Fungal Genet Biol 49:663-673. https://doi.org/10.1016/j .fab.2012.05.009.
- 38. Trail F, Wang Z, Stefanko K, Cubba C, Townsend JP. 2017. The ancestral levels of transcription and the evolution of sexual phenotypes in filamentous fungi. PLoS Genet 13:e1006867. https://doi.org/10.1371/ journal.pgen.1006867.
- 39. Cuomo CA, Güldener U, Xu JR, Trail F, Turgeon BG, Di Pietro A, Walton JD, Ma LJ, Baker SE, Rep M, Adam G, Antoniw J, Baldwin T, Calvo S, Chang YL, DeCaprio D, Gale LR, Gnerre S, Goswami RS, Hammond-Kosack K, Harris LJ, Hilburn K, Kennell JC, Kroken S, Magnuson JK, Mannhaupt G, Mauceli E, Mewes HW, Mitterbauer R, Muehlbauer G, Münsterkötter M, Nelson D, O'Donnell K, Ouellet T, Qi W, Quesneville H, Roncero MIG, Seong KY, Tetko IV, Urban M, Waalwijk C, Ward TJ, Yao J, Birren BW, Kistler HC. 2007. The Fusarium graminearum genome reveals a link between localized polymorphism and pathogen specialization. Science 317:1400-1402. https://doi.org/10.1126/science.1143708.
- 40. King R, Urban M, Hammond-Kosack MCU, Hassani-Pak K, Hammond-Kosack KE. 2015. The completed genome sequence of the pathogenic ascomycete fungus Fusarium graminearum. BMC Genomics 16:544. https://doi.org/10.1186/s12864-015-1756-1.
- 41. King R, Urban M, Hammond-Kosack KE. 2017. Annotation of Fusarium graminearum (PH-1) version 5.0. Genome Announc 5:e01479-16. https://doi.org/10.1128/genomeA.01479-16.
- 42. Son H, Seo YS, Min K, Park AR, Lee J, Jin JM, Lin Y, Cao P, Hong SY, Kim EK, Lee SH, Cho A, Lee S, Kim MG, Kim Y, Kim JE, Kim JC, Choi GJ, Yun SH, Lim JY, Kim M, Lee YH, Choi YD, Lee YW. 2011. A phenome-based functional analysis of transcription factors in the cereal head blight fungus, Fusarium graminearum. PLoS Pathog 7:e1002310. https://doi .org/10.1371/journal.ppat.1002310.
- 43. Wang C, Zhang S, Hou R, Zhao Z, Zheng Q, Xu Q, Zheng D, Wang G, Liu H, Gao X, Ma JW, Kistler HC, Kang Z, Xu JR. 2011. Functional analysis of the kinome of the wheat scab fungus Fusarium graminearum. PLoS Pathog 7:e1002460. https://doi.org/10.1371/journal.ppat.1002460.
- 44. Yun Y, Liu Z, Yin Y, Jiang J, Chen Y, Xu JR, Ma Z. 2015. Functional analysis of the Fusarium graminearum phosphatome. New Phytol 207: 119-134. https://doi.org/10.1111/nph.13374.
- 45. Kim HK, Jo SM, Kim GY, Kim DW, Kim YK, Yun SH. 2015. A large-scale functional analysis of putative target genes of mating-type loci provides insight into the regulation of sexual development of the cereal pathogen Fusarium graminearum. PLoS Genet 11:e1005486. https://doi .org/10.1371/journal.pgen.1005486.
- 46. Liu H, Wang Q, He Y, Chen L, Hao C, Jiang C, Li Y, Dai Y, Kang Z, Xu JR. 2016. Genome-wide A-to-I RNA editing in fungi independent of ADAR enzymes. Genome Res 26:499-509. https://doi.org/10.1101/gr.199877
- 47. Son H, Park AR, Lim JY, Shin C, Lee YW. 2017. Genome-wide exonic small interference RNA-mediated gene silencing regulates sexual reproduction in the homothallic fungus Fusarium graminearum. PLoS Genet 13:e1006595. https://doi.org/10.1371/journal.pgen.1006595.
- 48. Trail F, Common R. 2000. Perithecial development by Gibberella zeae: a light microscopy study. Mycologia 92:130-138. https://doi.org/10
- 49. Anavy L, Levin M, Khair S, Nakanishi N, Fernandez-Valverde SL, Degnan BM, Yanai I. 2014. BLIND ordering of large-scale transcriptomic developmental timecourses. Development 141:1161-1166. https://doi.org/
- 50. Guenther JC, Hallen-Adams HE, Bücking H, Shachar-Hill Y, Trail F. 2009. Triacylglyceride metabolism by Fusarium graminearum during colonization and sexual development on wheat. Mol Plant Microbe Interact 22:1492-1503. https://doi.org/10.1094/MPMI-22-12-1492.
- 51. Weirick T. Militello G. Müller R. John D. Dimmeler S. Uchida S. 2016. The identification and characterization of novel transcripts from RNA-seq data. Brief Bioinform 17:678-685. https://doi.org/10.1093/bib/bbv067.
- 52. Wang L, Park HJ, Dasari S, Wang S, Kocher JP, Li W. 2013. CPAT: Coding-Potential Assessment Tool using an alignment-free logistic regression model. Nucleic Acids Res 41:e74. https://doi.org/10.1093/nar/ gkt006.

Kim et al. mBio

 Chakraborty S, Deb A, Maji RK, Saha S, Ghosh Z. 2014. LncRBase: an enriched resource for lncRNA information. PLoS One 9:e108010. https:// doi.org/10.1371/journal.pone.0108010.

- Nam JW, Bartel DP. 2012. Long noncoding RNAs in *C. elegans*. Genome Res 22:2529–2540. https://doi.org/10.1101/gr.140475.112.
- Pauli A, Valen E, Lin MF, Garber M, Vastenhouw NL, Levin JZ, Fan L, Sandelin A, Rinn JL, Regev A, Schier AF. 2012. Systematic identification of long noncoding RNAs expressed during zebrafish embryogenesis. Genome Res 22:577–591. https://doi.org/10.1101/gr.133009.111.
- Li H, Wang Y, Chen M, Xiao P, Hu C, Zeng Z, Wang C, Wang J, Hu Z. 2016. Genome-wide long non-coding RNA screening, identification and characterization in a model microorganism *Chlamydomonas reinhardtii*. Sci Rep 6:34109. https://doi.org/10.1038/srep34109.
- Nyberg KG, Machado CA. 2016. Comparative expression dynamics of intergenic long noncoding RNAs in the genus *Drosophila*. Genome Biol Evol 8:1839–1858. https://doi.org/10.1093/gbe/evw116.
- Neil H, Malabat C, d'Aubenton-Carafa Y, Xu Z, Steinmetz LM, Jacquier A. 2009. Widespread bidirectional promoters are the major source of cryptic transcripts in yeast. Nature 457:1038–1042. https://doi.org/10 .1038/nature07747.
- Pelechano V, Wei W, Steinmetz LM. 2013. Extensive transcriptional heterogeneity revealed by isoform profiling. Nature 497:127–131. https://doi.org/10.1038/nature12121.
- Cabili MN, Trapnell C, Goff L, Koziol M, Tazon-Vega B, Regev A, Rinn JL. 2011. Integrative annotation of human large intergenic noncoding RNAs reveals global properties and specific subclasses. Genes Dev 25:1915–1927. https://doi.org/10.1101/gad.17446611.
- Ulitsky I, Shkumatava A, Jan CH, Sive H, Bartel DP. 2011. Conserved function of lincRNAs in vertebrate embryonic development despite rapid sequence evolution. Cell 147:1537–1550. https://doi.org/10.1016/ j.cell.2011.11.055.
- Motamedi MR, Verdel A, Colmenares SU, Gerber SA, Gygi SP, Moazed D. 2004. Two RNAi complexes, RITS and RDRC, physically interact and localize to noncoding centromeric RNAs. Cell 119:789 –802. https://doi.org/10.1016/j.cell.2004.11.034.
- Bühler M, Verdel A, Moazed D. 2006. Tethering RITS to a nascent transcript initiates RNAi- and heterochromatin-dependent gene silencing. Cell 125:873–886. https://doi.org/10.1016/j.cell.2006.04.025.
- Zhang K, Mosch K, Fischle W, Grewal SIS. 2008. Roles of the Clr4 methyltransferase complex in nucleation, spreading and maintenance of heterochromatin. Nat Struct Mol Biol 15:381–388. https://doi.org/10 .1038/nsmb.1406.
- Bayne EH, White SA, Kagansky A, Bijos DA, Sanchez-Pulido L, Hoe KL, Kim DU, Park HO, Ponting CP, Rappsilber J, Allshire RC. 2010. Stc1: a critical link between RNAi and chromatin modification required for heterochromatin integrity. Cell 140:666–677. https://doi.org/10.1016/j .cell.2010.01.038.
- Dang Y, Cheng J, Sun X, Zhou Z, Liu Y. 2016. Antisense transcription licenses nascent transcripts to mediate transcriptional gene silencing. Genes Dev 30:2417–2432. https://doi.org/10.1101/gad.285791.116.
- Park AR, Cho AR, Seo JA, Min K, Son H, Lee J, Choi GJ, Kim JC, Lee YW. 2012. Functional analyses of regulators of G protein signaling in *Gibberella zeae*. Fungal Genet Biol 49:511–520. https://doi.org/10.1016/j.fgb.2012.05.006.
- Chen Y, Gao Q, Huang M, Liu Y, Liu Z, Liu X, Ma Z. 2015. Characterization of RNA silencing components in the plant pathogenic fungus Fusarium graminearum. Sci Rep 5:12500. https://doi.org/10.1038/srep 12500.
- Mercer TR, Gerhardt DJ, Dinger ME, Crawford J, Trapnell C, Jeddeloh JA, Mattick JS, Rinn JL. 2011. Targeted RNA sequencing reveals the deep complexity of the human transcriptome. Nat Biotechnol 30:99–104. https://doi.org/10.1038/nbt.2024.
- Cabili MN, Dunagin MC, McClanahan PD, Biaesch A, Padovan-Merhar O, Regev A, Rinn JL, Raj A. 2015. Localization and abundance analysis of human IncRNAs at single-cell and single-molecule resolution. Genome Biol 16:20. https://doi.org/10.1186/s13059-015-0586-4.
- ENCODE Project Consortium. 2007. Identification and analysis of functional elements in 1% of the human genome by the ENCODE pilot project. Nature 447:799–816. https://doi.org/10.1038/nature05874.
- van Bakel H, Nislow C, Blencowe BJ, Hughes TR. 2010. Most "dark matter" transcripts are associated with known genes. PLoS Biol 8:e1000371. https://doi.org/10.1371/journal.pbio.1000371.
- Anderson DM, Anderson KM, Chang CL, Makarewich CA, Nelson BR, McAnally JR, Kasaragod P, Shelton JM, Liou J, Bassel-Duby R, Olson EN.

- 2015. A micropeptide encoded by a putative long noncoding RNA regulates muscle performance. Cell 160:595–606. https://doi.org/10.1016/j.cell.2015.01.009.
- Bánfai B, Jia H, Khatun J, Wood E, Risk B, Gundling WE, Kundaje A, Gunawardena HP, Yu Y, Xie L, Krajewski K, Strahl BD, Chen X, Bickel P, Giddings MC, Brown JB, Lipovich L. 2012. Long noncoding RNAs are rarely translated in two human cell lines. Genome Res 22:1646–1657. https://doi.org/10.1101/gr.134767.111.
- Gascoigne DK, Cheetham SW, Cattenoz PB, Clark MB, Amaral PP, Taft RJ, Wilhelm D, Dinger ME, Mattick JS. 2012. Pinstripe: a suite of programs for integrating transcriptomic and proteomic datasets identifies novel proteins and improves differentiation of protein-coding and noncoding genes. Bioinformatics 28:3042–3050. https://doi.org/10.1093/ bioinformatics/bts582.
- Wilson BA, Masel J. 2011. Putatively noncoding transcripts show extensive association with ribosomes. Genome Biol Evol 3:1245–1252. https://doi.org/10.1093/gbe/evr099.
- Chew GL, Pauli A, Rinn JL, Regev A, Schier AF, Valen E. 2013. Ribosome profiling reveals resemblance between long non-coding RNAs and 5' leaders of coding RNAs. Development 140:2828–2834. https://doi.org/ 10.1242/dev.098343.
- Ingolia NT, Brar GA, Stern-Ginossar N, Harris MS, Talhouarne GJS, Jackson SE, Wills MR, Weissman JS. 2014. Ribosome profiling reveals pervasive translation outside of annotated protein-coding genes. Cell Rep 8:1365–1379. https://doi.org/10.1016/j.celrep.2014.07.045.
- Juntawong P, Girke T, Bazin J, Bailey-Serres J. 2014. Translational dynamics revealed by genome-wide profiling of ribosome footprints in Arabidopsis. Proc Natl Acad Sci U S A 111:E203–E212. https://doi.org/ 10.1073/pnas.1317811111.
- Ji Z, Song R, Regev A, Struhl K. 2015. Many IncRNAs, 5'UTRs, and pseudogenes are translated and some are likely to express functional proteins. eLife 4:e08890. https://doi.org/10.7554/eLife.08890.
- Carlevaro-Fita J, Rahim A, Guigó R, Vardy LA, Johnson R. 2016. Cytoplasmic long noncoding RNAs are frequently bound to and degraded at ribosomes in human cells. RNA 22:867–882. https://doi.org/10.1261/ rna.053561.115.
- 82. Kang YJ, Yang DC, Kong L, Hou M, Meng YQ, Wei L, Gao G. 2017. CPC2: a fast and accurate coding potential calculator based on sequence intrinsic features. Nucleic Acids Res 45:W12–W16. https://doi.org/10.1093/nar/gkx428.
- Bogu GK, Vizán P, Stanton LW, Beato M, Croce LD, Marti-Renom MA. 2015. Chromatin and RNA maps reveal regulatory long noncoding RNAs in mouse. Mol Cell Biol 36:809–819. https://doi.org/10.1128/MCB .00955-15.
- 84. Wery M, Descrimes M, Vogt N, Dallongeville AS, Gautheret D, Morillon A. 2016. Nonsense-mediated decay restricts IncRNA levels in yeast unless blocked by double-stranded RNA structure. Mol Cell 61: 379–392. https://doi.org/10.1016/j.molcel.2015.12.020.
- Liu S, Wang Z, Chen D, Zhang B, Tian RR, Wu J, Zhang Y, Xu K, Yang LM, Cheng C, Ma J, Lv L, Zheng YT, Hu X, Zhang Y, Wang X, Li J. 2017. Annotation and cluster analysis of spatiotemporal- and sex-related IncRNA expression in rhesus macaque brain. Genome Res 27: 1608–1620. https://doi.org/10.1101/gr.217463.116.
- German MA, Pillay M, Jeong DH, Hetawal A, Luo S, Janardhanan P, Kannan V, Rymarquis LA, Nobuta K, German R, De Paoli ED, Lu C, Schroth G, Meyers BC, Green PJ. 2008. Global identification of microRNA-target RNA pairs by parallel analysis of RNA ends. Nat Biotechnol 26:941–946. https://doi.org/10.1038/nbt1417.
- Cloutier SC, Ma WK, Nguyen LT, Tran EJ. 2012. The DEAD box RNA helicase Dbp2 connects RNA quality control with repression of aberrant transcription. J Biol Chem 287:26155–26166. https://doi.org/10.1074/ jbc.M112.383075.
- Cheng P, He Q, He Q, Wang L, Liu Y. 2005. Regulation of the *Neurospora* circadian clock by an RNA helicase. Genes Dev 19:234–241. https://doi .org/10.1101/gad.1266805.
- Emerson JM, Bartholomai BM, Ringelberg CS, Baker SE, Loros JJ, Dunlap JC. 2015. period-1 encodes an ATP-dependent RNA helicase that influences nutritional compensation of the Neurospora circadian clock. Proc Natl Acad Sci U S A 112:15707–15712. https://doi.org/10.1073/pnas .1521918112.
- Zofall M, Yamanaka S, Reyes-Turcu FE, Zhang K, Rubin C, Grewal SIS. 2012. RNA elimination machinery targeting meiotic mRNAs promotes facultative heterochromatin formation. Science 335:96–100. https://doi .org/10.1126/science.1211651.

mBio[®]

- 91. Böhmdorfer G, Wierzbicki AT. 2015. Control of chromatin structure by long noncoding RNA. Trends Cell Biol 25:623-632. https://doi.org/10 .1016/j.tcb.2015.07.002.
- 92. Ni T, Tu K, Wang Z, Song S, Wu H, Xie B, Scott KC, Grewal SI, Gao Y, Zhu J. 2010. The prevalence and regulation of antisense transcripts in Schizosaccharomyces pombe. PLoS One 5:e15271. https://doi.org/10 .1371/journal.pone.0015271.
- 93. Rhind N, Chen Z, Yassour M, Thompson DA, Haas BJ, Habib N, Wapinski I, Roy S, Lin MF, Heiman DI, Young SK, Furuya K, Guo Y, Pidoux A, Chen HM, Robbertse B, Goldberg JM, Aoki K, Bayne EH, Berlin AM, Desjardins CA, Dobbs E, Dukaj L, Fan L, FitzGerald MG, French C, Gujja S, Hansen K, Keifenheim D, Levin JZ, Mosher RA, Müller CA, Pfiffner J, Priest M, Russ C, Smialowska A, Swoboda P, Sykes SM, Vaughn M, Vengrova S, Yoder R, Zeng Q, Allshire R, Baulcombe D, Birren BW, Brown W, Ekwall K, Kellis M, Leatherwood J, Levin H, Margalit H, Martienssen R, Nieduszynski CA, Spatafora JW, Friedman N, Dalgaard JZ, Baumann P, Niki H, Regev A, Nusbaum C. 2011. Comparative functional genomics of the fission yeasts. Science 332:930-936. https://doi.org/10.1126/science .1203357.
- 94. Zhang L, Ma H, Pugh BF. 2011. Stable and dynamic nucleosome states during a meiotic developmental process. Genome Res 21:875-884. https://doi.org/10.1101/gr.117465.110.
- 95. Chen HM, Rosebrock AP, Khan SR, Futcher B, Leatherwood JK. 2012. Repression of meiotic genes by antisense transcription and by Fkh2 transcription factor in Schizosaccharomyces pombe. PLoS One 7:e29917. https://doi.org/10.1371/journal.pone.0029917.
- 96. Cloutier SC, Wang S, Ma WK, Petell CJ, Tran EJ. 2013. Long noncoding RNAs promote transcriptional poising of inducible genes. PLoS Biol 11:e1001715. https://doi.org/10.1371/journal.pbio.1001715.
- 97. Cloutier SC, Wang S, Ma WK, Al Husini NA, Dhoondia Z, Ansari A, Pascuzzi PE, Tran EJ. 2016. Regulated formation of IncRNA-DNA hybrids enables faster transcriptional induction and environmental adaptation. Mol Cell 61:393-404. https://doi.org/10.1016/j.molcel.2015.12.024.
- 98. Ørom UA, Derrien T, Beringer M, Gumireddy K, Gardini A, Bussotti G, Lai F, Zytnicki M, Notredame C, Huang Q, Guigo R, Shiekhattar R. 2010. Long non-coding RNAs with enhancer-like function in human. Cell 143:46-58. https://doi.org/10.1016/j.cell.2010.09.001.
- Wang KC, Yang YW, Liu B, Sanyal A, Corces-Zimmerman R, Chen Y, Lajoie BR, Protacio A, Flynn RA, Gupta RA, Wysocka J, Lei M, Dekker J, Helms JA, Chang HY. 2011. A long noncoding RNA maintains active chromatin to coordinate homeotic gene expression. Nature 472: 120-124. https://doi.org/10.1038/nature09819.
- 100. Li W, Notani D, Ma Q, Tanasa B, Nunez E, Chen AY, Merkurjev D, Zhang J, Ohgi K, Song X, Oh S, Kim HS, Glass CK, Rosenfeld MG. 2013. Functional roles of enhancer RNAs for oestrogen-dependent transcriptional activation. Nature 498:516-520. https://doi.org/10.1038/nature 12210.
- 101. Boque-Sastre R, Soler M, Oliveira-Mateos C, Portela A, Moutinho C, Sayols S, Villanueva A, Esteller M, Guil S. 2015. Head-to-head antisense transcription and R-loop formation promotes transcriptional activation. Proc Natl Acad Sci U S A 112:5785-5790. https://doi.org/10.1073/pnas
- 102. Bitton DA, Grallert A, Scutt PJ, Yates T, Li Y, Bradford JR, Hey Y, Pepper SD, Hagan IM, Miller CJ. 2011. Programmed fluctuations in sense/ antisense transcript ratios drive sexual differentiation in S. pombe. Mol Syst Biol 7:559. https://doi.org/10.1038/msb.2011.90.
- 103. Bumgarner SL, Neuert G, Voight BF, Symbor-Nagrabska A, Grisafi P, van Oudenaarden A, Fink GR. 2012. Single-cell analysis reveals that noncoding RNAs contribute to clonal heterogeneity by modulating transcription factor recruitment. Mol Cell 45:470-482. https://doi.org/10 .1016/j.molcel.2011.11.029.
- 104. Rosa S, Duncan S, Dean C. 2016. Mutually exclusive sense-antisense transcription at FLC facilitates environmentally induced gene repression. Nat Commun 7:13031. https://doi.org/10.1038/ncomms13031.

- 105. Li N, Joska TM, Ruesch CE, Coster SJ, Belden WJ. 2015. The frequency natural antisense transcript first promotes, then represses, frequency gene expression via facultative heterochromatin. Proc Natl Acad Sci U S A 112:4357-4362. https://doi.org/10.1073/pnas.1406130112.
- 106. Metzenberg RL. 2004. Bird medium: an alternative to Vogel medium. Fungal Genet Rep 51:Article 8. https://doi.org/10.4148/1941-4765.1138.
- 107. Kim D, Langmead B, Salzberg SL. 2015. HISAT: a fast spliced aligner with low memory requirements. Nat Methods 12:357-360. https://doi.org/ 10.1038/nmeth.3317.
- 108. Pertea M, Pertea GM, Antonescu CM, Chang TC, Mendell JT, Salzberg SL. 2015. StringTie enables improved reconstruction of a transcriptome from RNA-seq reads. Nat Biotechnol 33:290-295. https://doi.org/10 .1038/nbt.3122.
- 109. Anders S, Pyl PT, Huber W. 2015. HTSeq—a python framework to work with high-throughput sequencing data. Bioinformatics 31:166-169. https://doi.org/10.1093/bioinformatics/btu638.
- 110. Robinson MD, McCarthy DJ, Smyth GK. 2010. edgeR: a Bioconductor package for differential expression analysis of digital gene expression data. Bioinformatics 26:139-140. https://doi.org/10.1093/bioinformatics/ htp616.
- 111. Law CW, Alhamdoosh M, Su S, Smyth GK, Ritchie ME. 2016. RNA-seq analysis is easy as 1-2-3 with limma, Glimma and edgeR. F1000Res 5:1408. https://doi.org/10.12688/f1000research.9005.2.
- 112. Bryant DM, Johnson K, DiTommaso T, Tickle T, Couger MB, Payzin-Dogru D, Lee TJ, Leigh ND, Kuo TH, Davis FG, Bateman J, Bryant S, Guzikowski AR, Tsai SL, Coyne S, Ye WW, Freeman RM, Jr, Peshkin L, Tabin CJ, Regev A, Haas BJ, Whited JL. 2017. A tissue-mapped axolotl de novo transcriptome enables identification of limb regeneration factors. Cell Rep 18:762-776. https://doi.org/10.1016/j.celrep.2016.12.063.
- 113. Aslett M, Wood V. 2006. Gene Ontology annotation status of the fission yeast genome: preliminary coverage approaches 100%. Yeast 23: 913-919. https://doi.org/10.1002/yea.1420.
- 114. Young MD, Wakefield MJ, Smyth GK, Oshlack A. 2010. Gene Ontology analysis for RNA-seq: accounting for selection bias. Genome Biol 11: R14. https://doi.org/10.1186/gb-2010-11-2-r14.
- 115. Wheeler TJ, Eddy SR. 2013. Nhmmer: DNA homology search with profile HMMs. Bioinformatics 29:2487-2489. https://doi.org/10.1093/bioinformatics/ btt403.
- 116. Langfelder P, Horvath S. 2008. WGCNA: an R package for weighted correlation network analysis. BMC Bioinformatics 9:559. https://doi.org/ 10.1186/1471-2105-9-559.
- 117. Zhang B, Horvath S. 2005. A general framework for weighted gene co-expression network analysis. Stat Appl Genet Mol Biol 4:Article 17. https://doi.org/10.2202/1544-6115.1128.
- 118. Lee HC, Li L, Gu W, Xue Z, Crosthwaite SK, Pertsemlidis A, Lewis ZA, Freitag M, Selker EU, Mello CC, Liu Y. 2010. Diverse pathways generate microRNA-like RNAs and Dicer-independent small interfering RNAs in fungi. Mol Cell 38:803-814. https://doi.org/10.1016/j.molcel.2010.04
- 119. Shahid S, Axtell MJ. 2014. Identification and annotation of small RNA genes using ShortStack. Methods 67:20-27. https://doi.org/10.1016/j .vmeth.2013.10.004.
- 120. Dichtl B, Stevens A, Tollervey D. 1997. Lithium toxicity in yeast is due to the inhibition of RNA processing enzymes. EMBO J 16:7184-7195. https://doi.org/10.1093/emboj/16.23.7184.
- 121. Frazee AC, Pertea G, Jaffe AE, Langmead B, Salzberg SL, Leek JT. 2015. Ballgown bridges the gap between transcriptome assembly and expression analysis. Nat Biotechnol 33:243–246. https://doi.org/10.1038/ nbt.3172.
- 122. Pertea M, Kim D, Pertea GM, Leek JT, Salzberg SL. 2016. Transcript-level expression analysis of RNA-seq experiments with HISAT, StringTie, and Ballgown. Nat Protoc 11:1650-1667. https://doi.org/10.1038/nprot.2016 .095.