Open reading frames provide a rich pool of potential natural antisense transcripts in fungal genomes

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Received July 29, 2005; Revised and Accepted August 15, 2005

ABSTRACT

Natural antisense transcripts are reported from all kingdoms of life and several recent reports of genomewide screens indicate that they are widely distributed. These transcripts seem to be involved in various biological functions and may govern the expression of their respective sense partner. Very little, however, is known about the degree of evolutionary conservation of antisense transcripts. Furthermore, none of the earlier analyses has studied whether antisense relationships are solely dual or involved in more complex relationships. Here we present a systematic screen for cis- and trans-located antisense transcripts based on open reading frames (ORFs) from five fungal species. The relative number of ORFs involved in antisense relationships varies greatly between the five species. In addition, other significant differences are found between the species, such as the mean length of the antisense region. The majority of trans-located antisense transcripts is found to be involved in complex relationships, resulting in highly connected networks. The analysis of the degree of evolutionary conservation of antisense transcripts shows that most antisense transcripts have no ortholog in any other species. An annotation of antisense transcripts based on Gene Ontology directs to common terms and shows that proteins of genes involved in antisense relationships preferentially localize to the nucleus with common functions in the regulation or maintenance of nucleic acids.

INTRODUCTION

Antisense transcripts are RNA sequences that are complementary to known (sense) transcripts. Experimental as well as in silico investigations have revealed substantial evidence that antisense transcripts are much more widespread than thought previously. However, it is still unclear to what extent antisense transcripts are generally functional. There are numerous examples of the detailed analysis of antisense transcripts functioning at the transcriptional and posttranscriptional level [for review see (1,2)] ranging from transcriptional control, splicing control, and degradative control of mRNA, up to higher order functions, such as gene silencing and imprinting (3,4). This implies that there is no single molecular mechanism of antisense function.

Several published investigations using different detection strategies report the occurrence of antisense transcripts in the human and mouse genomes. For example Lehner et al. (5) used pairwise BLAST analysis of curated sets of two mRNA libraries, while Shendure et al. (6) and Yelin et al. (7) used mRNA and Expressed sequence tag (EST)-libraries combined with information like exon–intron splicing structures and poly(A) signals to detect overlapping antisense transcripts from the same genomic locus.

Most analyses of antisense transcripts have focused on how widespread they are within one organism. The conservation across different species has been less studied, although many fundamental principles on the nature of antisense transcripts could be deduced from these studies. For instance, a comparison of antisense genes from human and mouse showed that less than half of all antisense genes between humans and mice have an ortholog in either species and that for only one-fifth of all pairs of antisense genes the antisense relationship is conserved (8).

To gain more insights into the abundance of antisense transcripts across eukaryotes, we decided to perform an analysis in four ascomycetes and one microsporidian fungi. A genomewide screening was performed in the genomes of the budding yeast Saccharomyces cerevisiae (9), its close relative Ashbya gossypii (10), the fission yeast Schizosaccharomyces pombe (11), the multicellular fungi Neurospora crassa (12) and the distant parasitic microsporidian Encephalitozoon cuniculi (13). At least some of these species are very diverged: orthologous genes between S.cerevisiae and S.pombe have on average the same distance between them as they have to their...
respective human orthologs (11). The species also show dramatic differences in organization of their genome: e.g. the number of genes with introns is greatly varying [S.cerevisiae (4%) (9) and S.pombe (43%) (11)]. They differ also in the capability to perform RNAi. There exist well known examples of RNAi in S.pombe, with important functions in the organization of heterochromatin (14,15) and cell division (16) and in N.crassa there are even two antisense systems (quelling and the classical RNAi) (17,18). However, there are no known examples of RNAi in S.cerevisiae, which lacks the main components of the RNAi machinery (19). Thus, it is of interest if these differences could influence the individual equipment with antisense transcripts in each species.

It has been shown both experimentally and computationally that antisense transcripts can also arise from trans-genomic loci (5,20). However, we are not aware of any systematic investigation of this phenomenon and, therefore, decided to perform a systematic search for both cis- and trans-located antisense transcripts based on nucleotide sequences of open reading frames (ORFs). We analyze the relationships of antisense transcripts leading to complex network structures, describe the degree of conservation across the species and functionally annotate the antisense transcripts based on the Gene Ontology (GO) vocabulary.

METHODS

Data sources

We used the annotated genomic sequences (NC_*) which are provided by the different fungal genome projects and which are redistributed by the RefSeq project (21) at NCBI. We downloaded the respective RefSeq releases of all five complete fungal genomes: S.cerevisiae (dated June 18, 2003), A.gossypii (dated March 10, 2004), S.pombe (dated January 9, 2004), N.crassa (dated January 22, 2004) and E.cuniculi (dated December 17, 2003). We extracted all sequence information for the ORFs as well as their annotation directly from the respective RefSeq files.

Antisense database

We used a self developed SQL-based database-schema (‘antisense-db’, unpublished data) suited to store and process antisense-related information. It is flexible and serves as a prototype implementation for the analysis of large sets of antisense transcripts. Virtually all analyses were performed by means of the ‘antisense-db’.

Identification of cis natural antisense transcripts (NATs)

The overlaps of cis NATs were determined directly from the genomic coordinates of ORF sequences that are specified in the respective RefSeq genome annotations. We report every antisense overlap occurring between ORFs with a minimal genomic overlap size of at least 1 bp. Since typical untranslated region (UTR) sizes are larger than 50 bp, the true overlap lengths are always larger than the size measured by the ORF overlap length.

Identification of trans NATs

To identify all ORFs that are antisense to each other, a pairwise BLAST-search using WU-BLASTN (W. Gish, personal communication) with the nucleotide sequences of all predicted ORFs was performed (E-value <10^-8). Besides the classical wobble-base pairs we also allowed the non-canonical basepairing of U-G. For that, we altered the nucleotide substitution matrices: the setting was M = 1, N = -2, Q = 3, R = 2, W = 9, wordmask = seg, lcmdist, V = 10000, B = 10000, E = 0.000001, -altscore = ‘C T 1’ and -altscore = ‘A G 1’. When more than one high-scoring pair (HSP) was reported, we combined their coordinates to one spanning region along the transcripts. For a given query sequence all HSPs from one subject sequence as reported by WU-BLAST were analyzed as follows: assume that there are n (overlapping) HSPs, such that i_k and j_k denote the first and last position, respectively, of the k-th HSP of the query, and i'_k and j'_k denote the first and last position, respectively, of the k-th HSP of the subject. Then take the HSP with the highest score, labeled m, and order all the remaining n-1 HSPs consistently with respect to their position, such that j_k < m < i_{k+1} and j'_k > m > i'_{k+1} for all k ∈ {1,2,...,n-1}. The longest consistent ordering is reported as the antisense overlap region of putative NAT partners. All cis NATs identified during the BLAST search are removed.

Identification of longest continuous stretch (LCS)-regions

The LCS is reported as the maximum length of matching base pairs (canonical and non-canonical basepairing U–G) in the antisense overlap regions of trans NATs.

Classification of overlap regions

The classification of overlap regions follows a simple schematic rule: we distinguish 5' overlaps, 3' overlaps and inside-overlaps. Any sequence was always considered from 5' to 3'. If an overlap occurs within the first quarter of the sequence and ends before the third quarter, we classify it as 5'-overlap. Vice versa, if an overlap occurs within the third quarter of the sequence and starts after the first quarter, we classify it as 3'-overlap. The remainders show no bias to one end of a sequence and are classified as inside-overlaps.

Detection of orthologous/paralogous sets

The determination of unique orthologs is not always possible due to the unequal expansion of protein family sizes (22) or the impact of genome duplication, e.g. as seen in S.cerevisiae (10). Therefore we decided to calculate sets of orthologs for an individual ORF. Thus we use a more relaxed definition of orthology, allowing for a specific ORF more than one orthologous partner in the other species [hence, we include also paralogs in our orthologous sets, similar to the definition of orthologous sets of paralogs given by Tatusov et al. (23) as used for the COG database]. The sets were identified first by using FastA (24) to compute pairwise alignments between the deduced protein sequences of the ORFs of each of the five species. We used a cutoff E-value of 10^-9 in all similarity searches. A given ORF was classified to have an ortholog in another species if it is ranked within the top three positions according to the E-value, and if its reciprocally reported ORF was also ranked within its best three hits. Thus, for a given ORF we allow maximally three orthologous ORFs in another species, and two ORFs from one species can share the same
ortholog. Let \( O_{ij} \), \( j \neq i \) be the relative number of ORFs in organism \( i \) that have an orthologous ORF in organism \( j \). Then we denote by \( O_{ij} \) the mean of all \( O_{ij} \), \( j = 1, \ldots, 5 \), \( j \neq i \). When we refer to the individual species, we use the letters \( C \) for \( S.cerevisiae \), \( A \) for \( A.gossypii \), \( P \) for \( S.pombe \), \( N \) for \( N.crassa \) and \( E \) for \( E.cuniculi \).

For the detection of paralogous (multigene) families within one species a similar approach was performed. Here, sets of ORFs were classified to be paralogous if every possible pair of ORFs is reported with an \( E \)-value \( < 10^{-9} \). In addition, to discriminate between members of multigene families and remote homologs a bidirectional overlap of at least 60% was required.

### Visualization of networks

To visualize the network of \( trans \) NATs we used yEd, a graphical editor included in the yFiles library (25). In the resulting graph a vertex represents an ORF involved in at least one antisense relationship and an edge is drawn between two vertices if the two corresponding ORFs form an antisense relationship.

### Annotation of NATs

The annotation of antisense transcripts is based on an ontology maintained by the GO consortium (26). The mapping from ontology terms to individual ORFs is performed by means of associaton files downloaded from the GO-website, which are available only for \( S.cerevisiae \) and \( S.pombe \) (http://www.geneontology.org). Since there are no association files provided for the other three species, orthology from \( S.cerevisiae \) ORFs to \( A.gossypii \), \( N.crassa \) and \( E.cuniculi \) ORFs was used to create the respective gene association files. A further refinement to gene associations was carried out by mapping the full ontology to GO-slim terms, a cut-down version of the full ontology containing a subset of the terms in the whole ontology. Then all common terms shared by NATs were detected using the GO-TermFinder perl modules (http://search.cpan.org/dist/GO-TermFinder/). This provides an object oriented set of libraries for dealing with files produced by the Gene Ontology project. From this analysis all significant common GO-terms with a \( P \)-value \( < 0.1 \) are reported. The \( P \)-values of a set of GO annotated genes are determined for a set of genes against the background of all genes in the genome sharing the same GO annotation. The \( P \)-value is calculated using the hypergeometric distribution as the probability of \( x \) or more out of \( n \) NAT ORFs having a given annotation, given that \( X \) of \( N \) (equal the total number of ORFs) have that annotation in the genome in general.

### RESULTS

#### Genomewide detection and classification of NATs

Our genomewide prediction of NATs is based on predicted ORFs. Analysis of the coding sequences of the five fungal genomes revealed a high number of ORFs exhibiting regions of base complementarity. Though the total number of antisense ORFs is varying to a great extent, we detected antisense ORFs in all five species (Table 1). The number of ORFs involved in antisense relationships ranges from 14.8% in \( S.cerevisiae \), 12.8% in \( E.cuniculi \), 6.4% in \( N.crassa \) to 2.5% in \( A.gossypii \) and 1.0% in \( S.pombe \) (compared with the total number of ORFs annotated in each genome).

We classify NAT pairs as \( cis \), if they populate the same genomic locus on the chromosomes, and as \( trans \), if they originate from different loci, with some ORFs having NAT partners belonging to both classes. \( cis \) NATs are directly calculated from their genomic coordinates. This allows the detection of \( cis \) NATs even for cases, where they exhibit very small overlap lengths of 1–20 bp, that are typically missed by heuristic alignment programs such as BLAST. We remind that with typical UTR sizes (often >50 bp), the true overlap lengths have to be larger than the size measured by the ORF overlap lengths. The ratio of detected \( cis \) NATs compared with the absolute number of ORFs in \( S.cerevisiae \) and \( E.cuniculi \) (with 11.3 and 9.1%) is up to 12 times larger than in the other three species with 1–2% \( cis \) NATs (Table 1).

We see no clear relationship between the gene density of the genomes and the number of \( cis \) NATs. While the maximal difference of gene density between any two species has a factor of 3.7 (\( E.cuniculi \) and \( N.crassa \)), the maximal density of \( cis \) NATs per kb genomic sequence varies up to a factor of 33 (\( E.cuniculi \) and \( N.crassa \)).

\( Trans \) NATs are also found in all five fungal organisms but again one observes large differences in the number of involved ORFs. While \(~5\%\) of all ORFs in \( S.cerevisiae \) (295 overall), \( N.crassa \) (561 overall) and \( E.cuniculi \) (100 overall) are \( trans \) NATs, only 0.2 (9 overall) and 0.6% (29 overall) of all ORFs are \( trans \) NATs in \( S.pombe \) and \( A.gossypii \), respectively.

### Table 1. Number of coding sequences (ORFs) involved in antisense relationships of five fungal organisms

|            | \( S.cerevisiae \) | \( A.gossypii \) | \( S.pombe \) | \( N.crassa \) | \( E.cuniculi \) |
|------------|---------------------|-----------------|--------------|----------------|-----------------|
| Total number of ORFs | 6304                | 4718             | 5041         | 10079          | 1996            |
| \( cis \) | 708 (11.2%)         | 91 (1.9%)        | 40 (0.8%)    | 87 (0.9%)      | 182 (9.1%)      |
| \( trans \) | 295 (4.7%)         | 29 (0.6%)        | 9 (0.2%)     | 561 (5.6%)     | 100 (5.0%)      |
| \( cis\)-\( trans \) | 73 (1.2%)          | 1 (0.02%)        | 1 (0.02%)    | 2 (0.02%)      | 26 (1.3%)       |
| NAT ORFs | 930 (14.8%)        | 119 (2.5%)       | 48 (1.0%)    | 646 (6.4%)     | 256 (12.8%)     |
| \( cis \) Pairs | 369 (26)           | 46 (0)           | 20 (0)       | 45 (0)         | 94 (16)         |
| \( trans \) Pairs | 411 (196)          | 19 (2)           | 8 (0)        | 722 (56)       | 235 (234)       |
| Combined | 780 (222)           | 65 (2)           | 28 (0)       | 767 (56)       | 329 (250)       |

The upper five rows of the table show the absolute number of ORFs involved in antisense relationships as well as relative number (compared with the total number of investigated coding-sequences). We distinguish \( cis \) antisense relationships, where both transcripts reside at the same genomic locus and \( trans \) antisense relationships where the ORFs originate from different loci. \( Cis\)-\( trans \) ORFs are involved in \( cis \) as well as \( trans \) relationships. NAT ORFs refer to the overall number of antisense ORFs. Therefore this number is equal to the sum of \( cis \) and \( trans \) NATs. The lower three rows of the table list the total number of antisense sequence pairs in each species. In parentheses the number of NAT pairs are listed where at least two ORFs from a multigene family are involved in antisense relationships (pairs of NATs).
Table 2. Characterization of antisense overlap region

| Species         | Identity | LOL (bp) | LCS (bp) | Corr(LOL/LCS) |
|-----------------|----------|----------|----------|---------------|
|                 | trans    | cis      | trans    |               |
| *S. cerevisiae* | 89% (8%) | 266.9 (151.9) | 190.6 (154.1) | 0.65*         |
| *A. gossypii*   | 78% (3%) | 88.9 (205.6) | 184.3 (64.1) | 0.01          |
| *S. pombe*     | 83% (4%) | 34.0 (36.5) | 132.8 (38.3) | 0.50          |
| *N. crassa*    | 84% (6%) | 218.9 (149.8) | 118.1 (52.3) | -0.44         |
| *E. cuniculi*  | 85% (8%) | 135.0 (150.3) | 45.3 (74.8) | 0.16          |

The table lists the mean identity of *trans* NATs, mean overlap-length (LOL) of *cis* and *trans* NATs, mean longest common stretch of identical residues (LCS) and Pearson correlation coefficient of overlap-length and LCS. (Numbers in parentheses indicate their respective standard deviation.)

*Subtracting all ORFs from multigene families (Table 1) results in a mean LOL = 124 bp, a mean LCS = 25 bp and a corr(LOL/LCS) = 0.16. Subtracting all ORFs from multigene families (Table 1) results in one remaining NAT pair with an LOL = 52 bp and an LCS = 52 bp.

Table 3. Genomic arrangement of NATs

| Overlap region | S. cerevisiae | A. gossypii | S. pombe | N. crassa | E. cuniculi |
|---------------|--------------|-------------|----------|-----------|-------------|
|               | *cis* | *trans* | *cis* | *trans* | *cis* | *trans* | *cis* | *trans* | *cis* | *trans* |
| 5'             | 319 (43%) | 337 (41%) | 37 (40%) | 19 (50%) | 4 (10%) | 2 (12%) | 36 (40%) | 213 (15%) | 130 (69%) | 215 (46%) |
| Inside         | 209 (28%) | 266 (32%) | 2 (2%) | 10 (26%) | 0 (0%) | 8 (50%) | 36 (40%) | 465 (32%) | 19 (10%) | 170 (36%) |
| 3'             | 210 (28%) | 219 (27%) | 53 (58%) | 9 (24%) | 36 (90%) | 6 (38%) | 18 (20%) | 766 (53%) | 39 (21%) | 85 (18%) |

Listed are the number of NATs (in parentheses relative to overall number of NATs) whose antisense region is located either 5', inside or 3' in transcript (see Methods for definition of 5', inside and 3' overlaps).
Orthologous antisense-transcripts

There are many examples of conserved functional entities in genomes, such as miRNAs or transcriptionally active regions, where the conservation is directly shown or at least believed to be due to functional constraints. Therefore we enquired to which extent ORFs involved in antisense relationships are conserved among the five fungal species.

To analyze conserved ORFs involved in antisense relationships we first determined the number of ORFs that have an orthologous sequence in any of the four other species. From this we computed the mean number of ORFs having orthologous sequences for every species. We notice that this varies by a factor of two between the species, ranging from 25.8% for O_N and 35.2% for O_E, to 50.3% for O_C, 50.8% for O_P, and 56.6% for O_A (for notation see Methods). The highest pairwise number was detected between the close relatives A. gossypii and S. cerevisiae: 88.9% of the ORFs from A. gossypii have an ortholog in S. cerevisiae.

Next we analyzed the degree of conservation of NAT pairs across the species. Here, we distinguished between pairs of NATs where both ORFs have orthologs in another species (BO-pair) and pairs of NATs where only one ORF has an ortholog in another species (LO-pair). Figure 1 illustrates that for many pairs of NATs only one of the two transcripts has an ortholog (i.e. are LO-pairs). For example, of 780 NAT pairs in S. cerevisiae, 493 have an ortholog: 453 are LO-pairs, while only 40 are BO-pairs. Exceptions are found for A. gossypii and for S. pombe where most orthologous NATs are BO-pairs (68 and 58%, respectively). For a complete list of all ortholog assignments for cis and trans NATs see Supplementary Table 2.

Evolutionary conserved NATs

We further enquired whether antisense relationships of ORFs are evolutionarily conserved. There are no NATs that are conserved as NATs in all species. We found two pairs of fully conserved NATs in S. cerevisiae and A. gossypii, and they are both at syntenic positions. In addition, we observed that the orthologs of NAT ORFs are rarely antisense transcripts themselves (Figure 1). The number of conserved cis NATs is generally low and we hypothesized that some cis NATs are still conserved as adjacent but non-overlapping ORFs on the chromosomes. To test this hypothesis, we determined the distance of adjacent orthologous ORFs. The majority of orthologs of cis NATs are >1000 bp apart from each other. For the cis NATs of A. gossypii, we found 14 syntenic regions, all in S. cerevisiae which are on average 266 bp apart. There are only two further cis NAT pairs, whose orthologous ORFs are <1000 bp apart from each other (one cis NAT in S. cerevisiae and one cis NAT in S. pombe with a syntenic region in A. gossypii).

Functional annotation of antisense transcripts

We used the annotations provided by the GO consortium to annotate all NATs from each species by means of the three classes of the ontology (function, process and component). Since only mappings of GO-terms to gene products for S. cerevisiae and S. pombe are available, we used orthology from ORFs of A. gossypii, N. crassa and E. cuniculi to the ORFs from S. cerevisiae to assign GO-terms to gene products in these species. Of 1059 S. cerevisiae ORFs which have an ortholog in E. cuniculi 594 ORFs are GO-annotated. Similarly, of 3304 N. crassa orthologs 2560 ORFs are GO-annotated and of 4937 A. gossypii orthologs 4361 ORFs have a GO-annotation. To compare the GO categories between each species, we mapped the full ontology to the GO slim terms provided by the Saccharomyces Genome Database (SGD) (http://www.yeastgenome.org). The GO slim terms are cut-down versions of the gene ontologies containing a subset of the terms of the whole ontology. They give a broader overview of the content without the detail of the specific fine grained terms.

The GO analyses of cis and trans NATs are summarized in Figures 2 and 3, respectively. To ensure significance we only list every GO term shared by ORFs having P-values <0.1 (Methods). Though there are specific annotations of GO terms for each species, there are also GO terms that are shared between species and even between cis and trans NATs. Almost all terms shared by more than one species are also shared between cis and trans NATs. Examples of the process ontology are the terms ‘nucleobase, nucleoside, nucleotide and nucleic acid metabolism’, which are in common by S. cerevisiae, S. pombe and N. crassa and by cis and trans NATs. Another example is the term ‘transcription’, annotated for S. cerevisiae, S. pombe and E. cuniculi and shared between cis and trans NATs at least in S. cerevisiae. From the annotation with the component ontology it can be seen that high numbers of NAT gene products localize to the nucleus, with 30–50% of the annotated trans NATs from S. cerevisiae, S. pombe and N. crassa, respectively, and 33 and 70% of the annotated cis NATs from S. cerevisiae and E. cuniculi, respectively. Removing any multigene-family-NAT does not significantly alter the results of the GO analysis. Changes are very specific, e.g. terms like ‘helicase activity’ no longer appear due to the removed helicases encoded by subtelomeric repeats in S. cerevisiae.
For cis NATs of *S. cerevisiae* significant functional groups are found. Here the biggest group consists of proteins that are involved in ribonucleoprotein complexes, with 32 participating ORFs (*P*-value < 0.1). An important subset of this group consists of 17 ORFs, all of them are constituents of the ribosome (*P*-value < 0.01). Highly abundant are also functional groups related to transcription factor activity (36 ORFs, *P*-value < 0.1) or chromosome organization (26 ORFs, *P*-value < 0.001).
Analysis of the antisense interaction network

We analyzed the network of *S.cerevisiae* NATs in detail. Most of the cis NATs have unique dual relationships (313 pairs), 28 cis ORFs have two different NAT partners and one has four different NAT partners. In contrast, trans NATs are involved in a much more complex network (Figure 4). There are many trans NATs involved in large networks of relationships, resulting in some bigger subgraphs (Figure 4A–C) and smaller subgraphs (Figure 4D–G) and some mostly dual relationships (Figure 4H). The subgraphs in Figure 4B–G are composed of NATs that are mostly members of multigene families, while the dual relationships (Figure 4H) and the largest subgraph (Figure 4A) contain almost only NATs that are devoid of multigene families.

As can be seen, some of these NATs have many antisense partners and they are connectors of the complex subgraphs. For example, the ORFs YOR053W and YBR113W have 52 and 31 trans NAT partners, respectively. We call these founders of antisense relations (FARs); if these elements are removed, the subgraphs disintegrate and many antisense relationships are abolished. The creation of FARs is often explained by the fact that many FARs overlap in antisense with one member from a larger protein family onto the same genomic locus (as a cis-relationship). Consequently, these FARs do also serve as founders for the complex trans-relationships with further members of the protein family.

An example of sense/antisense ORF transcription can be found in the subgraph (Figure 4E). As first reported by LeJohn (27,28) for the water mold *Achlya klebsiana* and later for *Drosophila auraria* (29), the genomic locus for the NAD-glutamate dehydrogenase harbors an antisense transcript which codes for a heat shock protein (HSP70).

Both transcripts give rise to proteins and examination via hydropathy plots reveals almost perfect mirror images between the deduced aminoacid sequences of the sense/antisense transcripts (27). The ORF YAL004W, described as having strong similarity to *A. klebsiana* glutamate dehydrogenase, has also a predicted cis-antisense transcript belonging to the HSP70 family (YAL005C), recapitulating the concept of FARs: it has also conserved trans-antisense relationships with further members of the HSP70 family.

The same principle holds for nearly all subnetworks found in *S.cerevisiae*. As noted above, two of the big networks and four of the small networks represent larger protein families (most of them are localized in subtelomeric regions). Most ORFs shown in Figure 4B are annotated Y'-elements, containing a helicase-encoding ORF which is expressed only during meiosis and in telomerase-deficient cells (30). The subgraph (Figure 4D) consists of proteins sharing the PAU domain, forming two families which are activated during anaerobic processes and involved in the formation of the cell wall (DAN family) (31) or active during alcoholic fermentation (Seripauperin family) (32).

The largest and most diverse network found for *S.cerevisiae* is shown in Figure 4A. It consists in total of 137 NATs and is mostly devoid of NATs from multigene families. The 137 NATs were analyzed by means of the GO-annotation as described above. Out of 137 NATs, 69 are annotated components of the ‘nucleus’. The translation products of 51 NATs are involved in ‘nucleobase, nucleoside, nucleotide and nucleic acid metabolism’, most of them are involved in ‘transcription’ (32 NATs) exhibiting ‘transcription regulator activity’ (28 NATs).

Analysis of the trans networks of the other four species remains challenging due to missing annotation. Most ORFs are annotated as ‘predicted proteins’ or ‘hypothetical proteins’. Nevertheless upon visual inspections many topological motifs are recognized, which are similar to those of the trans network of *S.cerevisiae*. The trans network of *N.crassa* is shown in Figure 5.

Transcribed NATs

Hurowitz *et al*. (33) analyzed almost all budding yeast ORFs by means of correlating their assumed transcript length with the individual measurements from microarray experiments, thus ensuring that every measurement on the array is due to a transcript of expected size (virtual northern technique). Of special interest are questionable ORFs. A list of 820 questionable ORFs, representing a curation of experimental and computational data from the following articles (34–39), was distributed by the SGD (http://www.yeastgenome.org). Hurowitz *et al*. confirmed the transcription of 192 ORFs of these ORFs, indicating that these 192 ORFs are likely to represent bona fide genes.

We compared the set of confirmed questionable ORFs of Hurowitz *et al*. with the set of *S.cerevisiae* ORFs involved in antisense relationships as predicted by our study. In addition, we used all unambiguous assignments of yeast ESTs [downloaded from SGD (ftp://genome-ftp.stanford.edu/pub/yeast/sachDB/)] to ORFs. Of 369 cis NAT pairs 152 NAT pairs are confirmed either by the virtual northern dataset of Hurowitz *et al*. or by the EST mapping or by direct
experimental evidence given by SGD. Correspondingly, of 
411 trans NAT pairs 206 NAT pairs are confirmed by either 
of the three data sources as aforementioned (for a detailed 
listing see Supplementary Table 3).

**DISCUSSION**

We have performed a genomewide screen of antisense tran-
scripts based on annotated ORFs in four ascomycete and one 
microsporidian fungi. The number of ORFs exhibiting anti-
sense relationships varies widely between all investigated 
species, even for the two closely related species *S.cerevisiae* 
and *A.gossypii*. This finding is remarkable considering the 
large number of orthologs between *A.gossypii* and *S.cerevisiae* 
and the strong synteny that is shown for both genomes (10). 
One explanation is that *A.gossypii* lacks any transposons and 
subtelomeric repeats, which are a major source of antisense 
transcripts in *S.cerevisiae*. But even when all effects of multi-
gen family or repetitive family are removed, there are 
profound differences between the genomes. The high variation 
in the total number of *cis* NATs in the five species contrasts 
with the general idea that the compactness of genomes influ-
ences the number of overlapping ORFs and shows that there is 
more than one factor affecting the total number of overlapping 
ORFs in a genome.

Several investigations of antisense transcripts in human 
and mouse have been published (6–8). The results of these 
analyses clearly depend on the number and kind of antisense 
transcripts found and the databases used in the mapping pro-
cedures. For example Shendure et al. (6) found 217 *cis* NAT 
pairs in human, while Yelin et al. (7) reported 2667 *cis* NAT 
pairs. Although the set of human antisense transcripts found 
by Yelin *et al.* contains also non-coding transcripts, they detec-
ted ~400 NATs with overlaps in coding regions (1.1% of 
30000 human genes). Strikingly we found similar numbers 
of *cis* NATs in *A.gossypii*, *S.pombe* and *N.crassa* (1.9, 0.8 and 
0.9%, respectively), but not in *S.cerevisiae* and *E.cuniculi*, 
where we found much higher numbers of *cis* NATs 
(11.3 and 9.1%, respectively). It is noteworthy that many 
previously overlooked ORFs in *S.cerevisiae* (40) show an 
unexpectedly high number of partial or complete overlaps 
with known ORFs, supporting the idea of the high occurrence 
of *cis* NATs in *S.cerevisiae*.

Though there are prokaryotic examples of *trans*-encoded 
antisense transcripts (41), not many examples of functional 
*trans* NATs in eukaryotes have been described. Generally, any 
requirements for functional *trans* NATs are still unknown. 
Rosok et al. (20) introduced a technical system to identify 
endogenous human mRNAs with long complementary regions 
to known transcripts, based on their capability to form stable 
RNA/RNA duplexes *in vitro*. Similarities of 90% in the com-
plementary regions were observed. This is the level of simi-
arity found by us in the overlap regions of *trans* NATs. 
Korneev *et al.* (42) described the *trans*-regulation of a neural 
nitric oxidase synthase (nNOS) by an antisense RNA tran-
scribed from an nNOS pseudogene. The overlap region of 
nNOS and its antisense RNA is 139 bp in length with a simi-
arity of 87%, and it has been shown that these two transcripts 
form stable duplexes *in vivo*. These reports show that NATs 
with overlap regions of 130 bp are capable of forming stable 
duplexes. Such overlap lengths and corresponding identities 
are detected for many *cis* NATs in *S.cerevisiae*, *E.cuniculi* and 
*N.crassa* as well as for *trans* NATs in *S.cerevisiae* and 
*E.cuniculi*, denoting them as possible candidates to form 
RNA/RNA duplexes *in vivo*.

Another important parameter for duplex formation is the 
accessibility of complementary sequences with respect to their 
own secondary and tertiary structures. The initial presenta-
tion of three to four bases suffices in many cases to provide 
the scaffold for rapid interaction with complementary RNA. 
For instance, the analysis of the ubiquitous YUNR RNA 
recognition motif showed that the first step in forming 
RNA/RNA-duplexes involves very few accessible nucleotides 
(43). Based on this observation we introduced the LCS as 
the longest chain of continuous matching base pairs in 
the antisense region, which is a potential target for duplex 
formation. The antisense overlap region of nNOS and its anti-
sense pseudogene exhibit an LCS of 18 bp (42). Antisense 
transcripts in *S.cerevisiae* and *E.cuniculi* exhibit even much 
longer LCS; their average LCS’ are ~60 and 45 bp, respect-
ively. Antisense transcripts in *N.crassa*, *A.gossypii* and 
*S.pombe*, on the other hand, have LCS of roughly one-third 
of those found in *S.cerevisiae* and *E.cuniculi*. This difference, 
however, is no longer existing when removing multigene-
family-NATs in *S.cerevisiae* and *E.cuniculi*. This, along 
with the strong anticorrelation of LCS in *N.crassa* and 
*S.pombe*, indicates, that there are potent molecular me-
chanisms to prevent longer, perfect duplexes in these two 
organisms.

In that respect it is interesting that there are many reports 
from *N.crassa* and *S.pombe* regarding RNAi, whereas it is 
known that *S.cerevisiae* has completely lost all known com-
ponents of RNAi (19). In *E.cuniculi* and *A.gossypii*, there is 
also no Dicer or Argonaute homolog encoded in the genome 
(S. Steigele, personal observation). It could be speculated that 
an organism with no endogenous response to double-stranded 
RNA (dsRNA), such as *S.cerevisiae* and *E.cuniculi*, tolerates 
more antisense transcripts by chance than an organism such as
that these NATs are remnants of earlier overprinting effects, trans NATs in cis reading frames. This is particularly interesting in the case of the nucleotide sequence is translated into different reading frames. This is particularly interesting in the case of cis NATs. For the majority of NATs only one NAT has an ortholog and many of the ORFan antisense partners could have been generated by virtue of the overprinting effects. Our analysis detected the highest number of conserved NATs between S. cerevisiae and N. crassa along with the interesting finding that many orthologs of cis NATs in S. cerevisiae are trans NATs in N. crassa and vice versa. It could be speculated that these NATs are remnants of earlier overprinting effects, with a pronounced selective pressure in N. crassa separating these ORFs.

Many ORFans are really transcribed, as shown for S. cerevisiae (50), and this suggests that these ORFans could be utilized to control the gene expression of their antisense partners at many levels (transcription, maturation, transport, stability and translation) (2). For example, the mRNAs of yeast are known to involve major control elements (51) which act through specific secondary (tertiary) RNA structures. There is an emerging number of RNA based regulation enlarging the ways of classical gene regulation. Isaacs et al. (52) showed that artificial riboregulators in the 5'-UTR of mRNAs are sensible to short complementary RNAs, which upon binding are competent to switch on/off the translation of the corresponding protein. A related mechanism acting on the stability of mRNAs is the interaction of short modifier RNAs with AU-rich elements (53). The basis of all these mechanisms are complementary RNAs perturbing or preserving a structural conformation of mRNA. It is conceivable that many antisense transcripts act in a transient fashion as modifiers of RNA secondary structures. One nice prediction of these models is that modifier RNAs evolve very fast and do not leave any phylogenetic footprints (54), perhaps one reason for the generally low conservation of antisense relationships we and others have seen.

The networks we describe have some interesting features in common. For instance, the trans NAT networks we described have some nodes with a high connectivity and many others with a low connectivity, similar to social and communication networks (55). One simple explanation for the emergence of the yeast network comes from the observation that the distribution of protein family sizes in yeast follows a power law (56), with a very small number of proteins having a large number of paralogs and vice versa. If many antisense relationships occur randomly in this set, it could explain the emergence of the complex network by simply displaying the underlying power-law distribution from the family sizes of the involved proteins. Nevertheless, this does not explain the high interconnectivity between each unit of the network, demanding for some yet unknown consolidating forces. The FARs in S. cerevisiae or E. cuniculi serve as an example. They have genomic overlaps with one member of a conserved protein family, resulting in trans relationships with further members of the protein family (multigene family). Although our model explains the networks from S. cerevisiae and E. cuniculi, it fails to explain the pronounced network of N. crassa, which is known to have a very low fraction of multigene families and no known repetitive elements.

In light of all these differences the functional annotation based on the GO revealed interesting results. We found significant groupings of specific terms between all species for both cis and trans NATs. Despite the diverse function of most NATs, many NATs are found to be related to aspects of nucleic acid metabolism and many of these proteins are involved in the regulation of transcription. Additionally, an overwhelming proportion of all proteins is specifically localizing to the nucleus, which was also recently found for human cis NATs (57). Lehner et al. (5) similarly categorized a high number of all human antisense transcripts to be involved in ‘nucleic acid binding’ and ‘transcription factor activity’.

In contrast to trans NATs, where most groups of ORFs sharing GO-terms are easily explained by virtue of the
FARs-effects, any functional grouping of ORFs related to *cis* NATs is remarkable. In *Saccharomyces* 32 proteins are involved in ‘ribonucleotide complexes’, comprising 17 proteins that are structural constituents of the ribosome and 11 proteins that are part of the ‘RNA-splicing’ machinery. This noticeable accumulation of functionally related NATs points to regulative constraints, which are important for the assembly and/or coordinated function of these proteins: in the case of ribosomes it is well known that the rate of accumulation of each ribosomal protein is carefully regulated by the yeast cell to provide the equimolar ratio necessary for the assembly of the ribosome (58).

Our analysis has shown that antisense transcripts are common in the species investigated in this study. They are less conserved than overlapping ORFs from bacteria. The functional characterization of antisense transcripts demonstrates an interesting accumulation of specific functional groups, as shown before for other species. The networks we have found have the potential to give important hints about the function and evolution of natural antisense regulated systems.

**SUPPLEMENTARY DATA**

Supplementary Data is available at NAR Online.

**ACKNOWLEDGEMENTS**

We thank Andreas Schwienhorst for reading and commenting an earlier version of this manuscript. We are also grateful to Peter Stadler and Stefanie Bette for discussions. We thank Peter Wills for reading and commenting the manuscript. This work was supported by the Deutsche Forschungsgemeinschaft, to Peter Stadler and Stefanie Bette for discussions. We thank Andreas Schwienhorst for reading and commenting the manuscript. This work was supported by the Deutsche Forschungsgemeinschaft, AZ BIZ 1/1-3. Funding to pay the Open Access publication charges for this article was provided by Deutsche Forschungsgemeinschaft.

*Conflict of interest statement.* None declared.

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