1	Transposon insertional mutagenesis in Saccharomyces uvarum reveals trans-acting
2	effects influencing species dependent essential genes
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4	Running Title
5	Gene essentially is influenced by trans effects
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1 Abstract

2	To understand how complex genetic networks perform and regulate diverse cellular
3	processes, the function of each individual component must be defined. Comprehensive
4	phenotypic studies of mutant alleles have been successful in model organisms in
5	determining what processes depend on the normal function of a gene. These results are
6	often translated to the increasing number of newly sequenced genomes by using sequence
7	homology. However, sequence similarity does not always mean identical function or
8	phenotype, suggesting that new methods are required to functionally annotate newly
9	sequenced species. We have implemented comparative functional analysis by high-
10	throughput experimental testing of gene dispensability in Saccharomyces uvarum, a sister
11	species of <i>S. cerevisiae.</i> We created haploid and heterozygous diploid Tn7 insertional
12	mutagenesis libraries in <i>S. uvarum</i> to identify species dependent essential genes, with the
13	goal of detecting genes with divergent function. Comprehensive gene dispensability
14	comparisons with <i>S. cerevisiae</i> revealed that approximately 12% of conserved orthologs are
15	predicted to display diverged dispensability, including 22 confirmed differentially essential
16	genes. Surprisingly, despite their differences in essentiality, these genes are capable of
17	cross-species complementation, demonstrating that other <i>trans</i> -acting factors that are
18	background dependent contribute to differential gene essentiality. Furthermore, we
19	identified an instance of swapped essentiality between two paralogs, CDC25 and SDC25
20	between these two species. This data set provides direct experimental evidence of gene
21	function across species, which can inform comparative genomic analyses, improve gene
22	annotation and be applied across a diverse set of microorganisms to further our
23	understanding of gene function evolution.

1 Keywords

- 2 Functional genomics, gene function evolution, yeast, insertional mutagenesis, essential
- 3 genes, orthologs

1 Introduction

2	The ability to accurately predict gene function based on DNA sequence similarity is
3	a valuable tool, especially in the current stage of genomic research where numerous
4	genomes are increasingly becoming sequenced. It has become crucially important to
5	predict gene function based on sequence similarity due to the lack of experimentally
6	determined functional information associated with each newly sequenced genome. Most
7	functional predictive methods rely on similarities of DNA sequence homology, co-
8	expression patterns as well as protein structure to help assign function to uncharacterized
9	genes, using genes where known functions have been previously characterized (Eisen 1998;
10	Usadel et al. 2009). However, these methods come with their own set of limitations and
11	often produce a substantial number of predictive errors, highlighting the importance of
12	implementing experimental methods to directly test gene function of previously
13	uncharacterized genomes to improve current methods of gene function annotation.
14	The gold standard of gene function characterization relies on targeted deletions of
15	predictive coding sequences to probe the contributions of each gene to specific biological
16	processes. To get a global view of gene function within an organism, several genome-wide
17	deletion collections have been created in model species, particularly in bacteria and yeast,
18	(Baba et al. 2006; Berardinis et al. 2008; Porwollik et al. 2014; Winzeler et al. 1999)
19	including highly diverged species (Kim et al. 2010; Schwarzmüller et al. 2014) as well as
20	different strains within a species (Dowell et al. 2010). These systematic deletion collections
21	are powerful tools for investigating molecular mechanisms of gene function, biological
22	pathways, and genetic interactions, especially in the genetic workhorse S. cerevisiae, where
23	gene function characterization and gene dispensability comparisons have been extensively
24	performed amongst various deletion collections of yeast (Costanzo 2016; Dowell et al.
25	2010; Kim et al. 2010; Tong et al. 2001). These studies have identified approximately 17%

1	of essential genes to be differentially essential between highly diverged species (S.
2	<i>cerevisiae</i> and <i>S. pombe</i>) and have discovered 6% of essential genes (57) that are
3	differentially essential even between two strains of <i>S. cerevisiae.</i>
4	However, considerable effort and resources are required to create these targeted,
5	systematic libraries and they are not a practical approach for validating gene function
6	across a wide range of non-standard genetic backgrounds in a high-throughput manner.
7	Alternative approaches to targeted gene deletion libraries are transposon based
8	mutagenesis methods used to create random insertional mutant collections, eliminating
9	requirements for a priori knowledge about defined coding regions and providing
10	information about partial loss of function or gain of function mutations. Random insertional
11	profiling has been widely applied across various species and has been instrumental in
12	understanding virulence genes, stress tolerance mechanisms and even tumor suppressor
13	genes in mice (DeNicola et al. 2015; van Opijnen and Camilli 2013; de la Rosa et al. 2017;
14	Weerdenburg et al. 2015; Yung et al. 2015; Coradetti et al. 2017). Several transposon
15	libraries have also been implemented across diverged yeast species, providing useful
16	information about gene function, growth inhibiting compounds and essential functional
17	protein domains (Gangadharan et al. 2010; Guo et al. 2013b; Michel et al. 2017; Oh et al.
18	2010; Ross-Macdonald et al. 1999; Zhao et al. 2017; Price et al. 2016) .
19	Here we utilize a random insertional method that has allowed us to assay gene
20	dispensability of approximately 50,000 mutants in Saccharomyces uvarum, a species that
21	diverged from S. cerevisiae approximately 20 million years ago and contains approximately
22	80% identity in coding sequences to <i>S. cerevisiae</i> (Dujon 2010; Kellis et al. 2003; Scannell et
23	al. 2007). These species can inter-mate to create hybrids, allowing us to leverage the large
24	genetic toolsets established in <i>S. cerevisiae</i> to more fully explore the genetic basis for
25	possible differential gene dispensability among these species. Genes with different

1	dispensability patterns between these two species can be used as a preliminary indicator of
2	divergent gene function, providing a model for investigating gene function evolution
3	between two diverged species of yeast. While this Tn7 insertional density is modest in
4	comparison to other insertional mutant libraries (50,000 compared to > 300,000
5	insertions) we successfully validated a subset of predicted differentially essential genes,
6	proving this approach to be useful for prioritizing genes for testing viability (Michel et al.
7	2017; Guo et al. 2013a). Furthermore, this Tn7 transposon mutagenesis library provides a
8	valuable resource for studying <i>S. uvarum</i> gene function and serves as a framework for
9	comparative functional genomics studies across newly sequenced, previously
10	uncharacterized species.
11	
12	Results
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13 14	Generating Tn7 insertional libraries in <i>S. uvarum</i> to predict essential and
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1 transposition of the Tn7 transposon was performed in a plasmid library containing random 2 *S. uvarum* genomic fragments. The Tn7 transposon was designed to carry a ClonNat 3 resistance marker that carries stop codons in all reading frames near both termini. The 4 interrupted genomic fragments were excised out of the plasmid and integrated at their 5 corresponding genomic positions, likely producing truncations when inserted within coding 6 regions (**Supplemental Fig. 1A**). The plasmid library contains ~50,000 unique genomic 7 insertion sites that were integrated into a diploid and a haploid *MATa* strain at a 10X 8 coverage (additional details can be found in **Supplemental Information**). Pools of mutants 9 from each Tn7 library were grown up in liquid cultures as described in Materials and 10 Methods. Insertion sites were determined using sequencing methods as described in detail 11 in the **Supplemental Information** along with DNA sequencing library preparation 12 protocols (Supplemental Fig. 1C). 13 The distribution of haploid, diploid and overlapping (present in both libraries) 14 insertion sites are evenly distributed throughout the *S. uvarum* genome, as illustrated in 15 **Supplemental Fig. 2.** (Detailed information about overall sequencing coverage is listed in 16 **Supplementary File 4.**) Once the insertion sites were determined in both libraries, we 17 counted the number of insertion sites in each annotated open reading frame using a custom 18 Python script (Materials and Methods). **Supplemental Table 1** summarizes the number of 19 insertion sites and the number of genes that contain insertion sites within each library, 20 including the initial plasmid library. All annotated S. uvarum genes containing the number 21 of insertion sites from each library are listed in **Supplementary File 5.** Of the 5,908 22 annotated genes, a total of 5,315 (90%) genes harbor insertion sites that were identified in 23 at least one library. Comparisons between shared genes and unique genes harboring 24 insertion sites are illustrated in **Supplemental Fig. 3**.

1 Since the essentiality of most genes is expected to be conserved between *S*. 2 cerevisiae and S. uvarum, we used the known essential set in S. cerevisiae to test if essential 3 genes in the haploid library contain fewer insertion sites. We identified a significant 4 reduction in the number of inserts present in known S. cerevisiae essential genes in the haploid library (Wilcoxon test $p < 2.2 e^{-16}$, essential average inserts/kb=0.88, SD=1.28 vs. 5 non-essential average inserts/kb = 4, SD = 4.38) (Supplemental Fig. 4), indicating that 6 7 essential genes are effectively targeted by this approach. However, due to the nature of the 8 library, insertional events at different positions across a gene may result in a partial loss of 9 function. Since essential genes may still tolerate some insertions, we instead relied on 10 comparisons between the diploid and haploid libraries to make inferences about gene 11 essentiality. Specifically, we calculated an insertion ratio using the number of inserts per 12 gene in the haploid library divided by the number of inserts in the diploid library, which 13 inherently normalizes for the length of the gene (Materials and Methods). Using the 14 insertion ratio as a metric, we also identified a significant difference between S. uvarum 15 genes whose orthologs are known to be essential (Wilcoxon rank tests Sc E:Sc NE p < p2.2e⁻¹⁶) (Supplemental Fig. 5B). Using the insertion ratio, we categorized genes as 16 17 essential or non-essential using a null distribution to rank genes above or below a cut-off 18 metric of 0.25 (details described in **Supplemental Information**). Using this cut-off 19 value, 1170 genes were categorized as essential genes. We applied an additional cut-off 20 metric (more details in Material and Methods) to remove a class of low coverage genes, 21 resulting in a total number of 718 (13%) predicted essential genes and 3,838 (65%) genes 22 that are predicted non-essential, with 1299 genes (22%) undetermined (genes without 23 inserts in the diploid library). We proceeded to characterize each gene set and validate 24 the dispensability of each of the predicted gene categories.

1

2 Analysis of predicted gene dispensability

3 The predicted gene list of *S. uvarum* essential genes was compared to known 4 essential genes lists from both S. cerevisiae and S. pombe to determine the amount of 5 conservation that exists between orthologs across diverged species. Of the predicted 718 S. 6 uvarum essential genes, 297 genes (42%) are shared amongst all three sets, with a total of 7 487 genes (68%) shared with at least one other set. Furthermore, 9 genes that are S. 8 cerevisiae strain specific, including 4 genes that are S288C specific and 5 genes that are 9 Σ 1278b specific, overlap with *S. uvarum* predicted essential genes (**Supplemental Fig. 6**). 10 Similar to what has been previously shown in S. cerevisiae, predicted essential genes in S. 11 *uvarum* were more likely to be unique, with 91% of essential genes (656/718) being 12 present in single copy compared to 76% of non-essential genes (2736/3604). Additionally, 13 comparisons between Gene Ontology (GO) molecular function terms of essential gene sets 14 from both species show significant enrichment (p-value < 0.01) for fundamental biological 15 functions. Processes such as DNA replication/binding, RNA and protein biosynthesis, as 16 well as structural constituents of the ribosome and cytoskeleton were enriched in both sets 17 of essential genes (Supplemental File 6). In contrast, non-essential genes were 18 significantly (p-value < 0.01) enriched for regulatory functions (transcription factor 19 activity) and conditional responsive processes, such as transmembrane transporter activity 20 and cell signaling (kinase activity) (Supplemental File 7). 21 Once we determined that many of the features of the predicted essential genes 22 were similar to confirmed essential genes in other species, we proceeded to create 23 heterozygous deletions to validate 13 conserved essential genes. Sporulating each 24 heterozygous deletion strain and performing tetrad analysis for cell viability confirmed 25 essentiality for 12 (92%) of the 13 strains (Supplemental Table 2). One example of a

1	confirmed essential gene can be found in Figure 1A , which illustrates the genomic positions
2	of all insertion sites across a genomic locus of chromosome V that contains essential and
3	non-essential genes. The color of the gene outline matches the predicted dispensability,
4	which is determined by their insertion ratio. For example, the gene <i>BRR2</i> has an insertion
5	ratio of 0.130 and is predicted to be a conserved essential gene (Figure 1C). The tetrad
6	analysis of a <i>BRR2</i> heterozygous deletion strain displays a 2 viable:2 inviable segregation
7	pattern in both species, validating this gene as a conserved essential gene (Figure 1B).
8	Images of all other confirmed essential genes are located in Supplemental Fig. 7. We also
9	tested three conserved non-essential genes and all three (100%) were confirmed as non-
10	essential (Supplemental Fig. 8) (Supplemental Table 2). Additionally, we obtained an
11	independent set of haploid deletion strains (see Materials and Methods), which was used as
12	a validated non-essential gene set. Out of the total 356 gene deletions that were included in
13	our library, 346 of those genes were predicted to be non-essential (97%) while the
14	remaining 3% were predicted to be essential indicating that a significant proportion of our
15	predicted non-essential genes were correctly predicted.





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14 Gene dispensability comparisons of orthologous pairs between *S. cerevisiae* and *S.*

15 uvarum

Our main goal of this project is to identify genes with differential essentiality to test
for evidence of divergent function. While the previous section categorized all annotated *S*. *uvarum* genes, we narrowed our analysis to 4,543 orthologous genes for which we had data
in the *S. uvarum* dataset to make direct comparisons of dispensability between *S. cerevisiae*

1 and S. uvarum (Supplementary File 8). Overall, 88% (4016/4543) of these genes display 2 conserved dispensability between S. cerevisiae and S. uvarum. The remaining 12% of 3 orthologs are predicted to differ in essentiality between the two species, with 304 (7%) of 4 these genes only essential in *S. uvarum* and 221 (5%) genes only essential in *S. cerevisiae* 5 (Supplemental Figure 9). Note the larger number of predicted essential genes in S. 6 uvarum (304 in S. uvarum compared to 222 in S. cerevisiae). This difference may be 7 attributed to the reliance on the absence of data (lack of insertion sites in a haploid gene) in 8 a greater proportion of genes not previously characterized as an essential gene in S. 9 cerevisiae (3687 non-essential genes vs. 765 essential genes), whereas, the latter category 10 utilizes the presence of insertional data in a smaller proportion of genes that are known to 11 be essential in *S. cerevisiae*. All predicted genes that differ in dispensability are listed in 12 Supplemental File 9. 13 To analyze the two categories of genes that differ in essentiality further, we 14 compiled a list of 222 genes from the S. cerevisiae-specific category and a more restrictive 15 list (Materials and Methods) of 220 *S. uvarum*-specific genes to normalize the number of 16 genes from each species. Using this list, we determined the proportion of *S. cerevisiae*-17 specific and S. uvarum-specific genes annotated for each function by performing Gene 18 Ontology (GO) term finder using the molecular function ontology. The proportion of 19 essential genes that differ between species for each functional category are represented in 20 **Supplemental Figure 9B,** illustrating a subset of all significant functional categories. 21 Interestingly, the most striking difference is in the functional category of the structural 22 constituent of the ribosome. This category is enriched for genes that are predicted to be 23 essential in *S. uvarum* (46/52). Additionally, differences exist between essential genes in 24 the category of RNA polymerase activity, where 9/10 genes were identified from predicted

1	<i>S. cerevisiae</i> essential genes. Full lists of significant (p-value < 0.01) GO enrichment
2	molecular function terms for each species individually are listed in Supplemental File 10 .
3	To confirm a subset of these predicted essential genes within each genetic
4	background, we applied the same method previously described to confirm conserved
5	essential genes by sporulating heterozygous deletion strains to determine the viability
6	pattern of the segregants. For example, gene SSQ1 is an example of a confirmed S. uvarum-
7	specific essential gene, illustrated in Figure 2 . An example of one confirmed <i>S. cerevisiae</i> -
8	specific gene is <i>VTC4</i> , illustrated in Figure 3 . Overall, we confirmed a total of 22 predicted <i>S</i> .
9	uvarum-specific and S. cerevisiae-specific essential genes (tetrad analysis can be found in
10	Supplemental Figs. 10 and 11 respectively). Interestingly, there are a variety of growth
11	phenotypes associated with confirmed species-specific genes. All combined tetrad analysis
12	results from the confirmation tests, also including false positives, are represented in
13	Supplemental Fig. 12. Supplemental Table 2 summarizes the total number of genes
14	confirmed in each category. We note the higher false positive rate in the species-specific
15	essential gene categories and attribute this to the low density of the overall library
16	coverage.









2 Figure 3. Validation of *S. cerevisiae*-specific essential gene *VTC4*. A) Mapped

3 chromosomal insertion positions are plotted across chromosome XII. Haploid inserts are

4 indicated in red, diploid inserts are blue and overlapping inserts are indicated in purple.

5 Genes indicated across the top are outlined according to predicted dispensability and filled

6 in if confirmed. Light pink filling indicates a gene that is essential in *S. cerevisiae* and non-

7 essential in *S. uvarum* (confirmed NE_E). B) Tetrad analysis of a heterozygous *vct4*Δ strain

8 displaying viable segregants with the $vtc4\Delta$ allele in *S. uvarum* plated on YPD and G418. C)

9 Tetrad analysis of a heterozygous *vtc4*^Δ allele in *S. cerevisiae* resulting in inviable

10 segregants. D) Table indicating the insertion ratio (number of haploid inserts by the

11 number of diploid inserts) per gene. The final column lists the predicted classification (NE,

- 12 non-essential; E, essential; N/A, no data).
- 13

14

In some cases, validation failures could be explained by errors in genome

15 annotation, not errors in our insertional library results. One gene, *DRE2*, was predicted to

16 be a non-essential gene but was confirmed as an essential gene through tetrad analysis.

17 Although this gene was initially called as a false positive, closer analysis revealed that all the

- 18 haploid insertions were clustered at the 5' end of the gene (Supplemental Fig. 13).
- 19 Performing protein alignments of *DRE2* between *S. cerevisiae* and *S. uvarum* revealed an

1	annotated start codon in <i>S. uvarum</i> upstream of the annotated start codon in <i>S. cerevisiae.</i> It
2	is likely that the gene was misannotated in <i>S. uvarum</i> , and instead shares the methionine
3	start position further downstream. Using the reannotated gene coordinates, we would
1	comential description DDE2 as accountial in Concerning since the bandoid in continuous as longer
4	correctly classify <i>DRE2</i> as essential in <i>S. uvurum</i> since the haploid insertions were no longer
5	included in the open reading frame. This one example highlights the utility of this library to
6	improve gene annotation in addition to characterizing gene essentiality.
7	To determine if the differences in dispensability could be explained by gene
8	expression, we compared a previously described metric used to quantify differences in gene
9	expression between orthologous genes. We compared this metric in known genes that
10	differ in essentiality and did not find evidence of genes enriched for expression differences,
11	suggesting that gene expression alone cannot account for the differences in species
12	dependent essentiality (Supplemental Figure 14).
13	
13 14	Paralog divergence and duplicate gene loss explain some background effects on
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1	complementation assays by cloning <i>S. uvarum</i> alleles of both paralogs into a CEN/ARS
2	plasmid and testing whether the <i>S. uvarum</i> alleles could rescue the inviable phenotype of
3	segregants from a heterozygous $cdc25\Delta$ deletion in <i>S. cerevisiae</i> . We found that <i>SDC25</i> from
4	S. uvarum is functional and that both SDC25 and CDC25 alleles from S. uvarum can
5	complement a <i>cdc25</i> ⁴ deletion in <i>S. cerevisiae</i> (Table 1). Although we did not test for
6	complementation in the S. uvarum background, the results from the complementation
7	assays suggest that perhaps CDC25 is required for growth in S. cerevisiae due to the lack of
8	redundancy as a consequence of the non-functional copy of <i>SDC25</i> .
9	Following this same logic, we hypothesized that <i>CDC25</i> non-essentiality in <i>S</i> .
10	<i>uvarum</i> could be attributed to the redundancy provided by the functional copy of <i>SDC25</i> in
11	this species. To test this idea, we created an <i>S. uvarum</i> mutant heterozygous for both
12	$cdc25\Delta$ $sdc25\Delta$ and performed segregation analysis on the dissected tetrads (Table 1).
13	Unexpectedly, the segregation pattern of a double mutant displays a lethal phenotype for
14	not only the double mutant but also the single $sdc25\Delta$ mutant. We confirmed this result by
15	constructing an <i>sdc25</i> Δ heterozygous mutant in <i>S. uvarum</i> and found a 2:2 segregation
16	pattern showing that <i>SDC25</i> is an essential gene in <i>S. uvarum</i> . Although there is no clear
17	explanation for the requirement of SDC25 in S. uvarum, this comparison displays one clear
18	example of paralog divergence in essentiality between these two species.

Genotype	S. cerevisiae	S. uvarum
cdc25∆	-	+
sdc25∆	+	-
cdc25∆ + ScCDC25	+	N/D
cdc25∆ + SuCDC25	+	N/D
cdc25∆ + SuSDC25	+	N/D
$cdc25\Delta$ + ScSDC25	-	N/D
$sdc25\Delta + ScCDC25$	N/D	-

1

Table 1. Viability summary of gene deletions and complementation assays. Signs in
columns indicated by each species represent viability (-:inviable, +:viable, N/D:not done).
Complementation assays are represented by the gene deletion with the addition of each
gene expressed on a low copy plasmid.

6 7

8 In addition to redundancy differences that are attributed to divergence between 9 pairs of paralogs, gene gains and losses may also contribute to genetic background effects 10 that result in differential dispensability between *S. cerevisiae* and *S. uvarum*. Previous 11 studies have investigated gene gains and losses across the Saccharomyces clade and 12 identified genes that lost their duplicate in some species but not others. For example, ALR1 13 is found as a singleton in *S. cerevisiae* but has retained the duplicate copy in *S. uvarum*. The 14 ALR1 gene is a confirmed S. cerevisiae-specific essential gene, which may be explained by 15 the loss of the other copy of the duplicate pair. We created a heterozygous deletion of both 16 ALR1 copies in S. uvarum and tested the viability phenotypes of each mutant and of the 17 double mutant separately. Surprisingly, each single mutant and the double mutant were 18 viable. Furthermore, *ALR2* (the paralog of *ALR1*) is also nonessential, and even the triple 19 mutants had no phenotype (Supplemental Fig. 15).

20

1 Divergent gene dispensability is largely due to trans effects

2	While genetic redundancy or gene loss is a possible explanation for a fraction of
3	differentially essential genes, the remaining much larger portion of this class of genes
4	remained unexplained. Because our main goal for this study was to find evidence of gene
5	function divergence between these two species, we proceeded to further investigate the
6	remaining differentially essential genes for functional differences. For a subset of these
7	genes, we performed complementation assays in both species to test for divergent function.
8	We cloned five <i>S. cerevisiae</i> alleles from the list of <i>S. uvarum</i> -specific essential genes (<i>SAC3</i> ,
9	TUP1, CCM1, SSQ1, and AFT1) and seven S. uvarum alleles from the list of S. cerevisiae-
10	specific essential genes (ALR1, SHR3, CDC25, INN1, LCD1, SEC24, VTC24) into a CEN/ARS
11	plasmid to perform complementation tests in <i>S. uvarum</i> and <i>S. cerevisiae</i> (Supplemental
12	Fig. 16). The results from these complementation tests revealed that all genes are able to
13	complement the inviable phenotype in the other species, suggesting that the differences in
14	essentiality are more likely to be due to <i>trans</i> -acting changes rather than functional
15	differences of protein coding regions.
16	
17	Discussion
18	In this study, we applied a comparative functional genomics approach to investigate
19	how genetic background influences gene dispensability between two diverged species of
20	yeast. Using insertional integration comparisons between haploid and diploid pools of
21	mutants, we prioritized genes to validate as predicted essential, non-essential and
22	differentially essential gene categories in <i>S. uvarum.</i> We predicted approximately 12% of
23	orthologs to differ in dispensability between <i>S. uvarum</i> and <i>S. cerevisiae</i> and validated 22
24	genes in this category. Surprisingly, however, most genes that differ in dispensability have
25	retained their function between these two species, suggesting that differences in gene

1 dispensability are likely due to *trans*-acting changes rather than the direct result of

2 divergent coding sequence.

3 Specifically, our comparison of orthologous genes between *S. cerevisiae* and *S.* 4 *uvarum* revealed that a majority of genes maintain conserved dispensability requirements 5 (88%) while 12% of orthologs are predicted to be essential in one species but not the other. 6 We confirmed 93% (15/16) of predicted conserved categories of essentiality and 49% 7 (27/55) of genes predicted to be differentially essential. Although our rate of confirmed 8 genes in this category was lower than the conserved category, we correctly identified a 9 subset of genes that are differentially dispensable, despite the moderately dense insertional 10 profile of the library and a less restrictive cut-off value applied to include more genes to be 11 classified as this type. Further analysis of predicted species-specific essential genes 12 revealed enriched GO ontology terms of molecular functions involved in structural 13 constituent of the ribosome and DNA binding, although more precise analysis of functional 14 enrichments may require more thorough validation to remove the influence of false 15 positives. Finally, we utilized genetic tools in *S. uvarum* to test hypotheses about genetic 16 background effects that contribute to differences in essentiality. We find that differences 17 can be explained by paralog divergence and *trans*-acting factors. 18 Applying a random insertional approach has proved to be useful in functionally 19 profiling *S. uvarum* and will be useful for studying other understudied species, with the goal 20 of adding information to gene annotation methods. While this study was performed in

21 standard laboratory conditions, it is easily amenable for testing stressful conditions, other

22 nutrient sources as well as naturally relevant conditions. This library can be applied to

23 probe previously un-annotated genes or even proto-genes for functional acquisition, since it

24 is not restricted to *a priori* assumptions of genic boundaries. The identification of synthetic

25 lethal interactions can also be determined by performing insertional profiling in the

1	background of a particular mutation of interest relatively quickly and economically.
2	Additionally, pooled competition experiments en masse can be used to determine the
3	frequency of particular insertional mutants, providing quantitative measurements of
4	cellular fitness across conditions. Such a strategy could be efficiently employed using
5	computational approaches to prioritize experimental conditions that are most likely to
6	probe the most valuable phenotypic information for further functional characterization
7	(Guan et al. 2010).
8	Gene regulation also plays a large role in evolution and is crucial for responding to
9	environmental change (Carroll 2005). In previous studies, we aimed to functionally
10	characterize differences in gene expression patterns between S. cerevisiae and S. uvarum
11	and discovered species-specific responses to osmotic stress, peroxisome biogenesis and
12	autophagy, suggesting that each species may been exposed to different selective pressures
13	within their respective evolution histories (Caudy et al. 2013; Guan et al. 2013).
14	Interestingly, we did not find that genes with different gene expression patterns between
15	species were more likely to be differentially essential. Instead, trans genetic interactions
16	dominate. Identifying the molecular basis of these <i>trans</i> effects can now be undertaken,
17	potentially revealing principles of genetic interactions across species.
18	

1 Materials and Methods

2

3 Strains, plasmids and primers

4 The strains, plasmids and primers used in this study are listed in **Supplementary** 5 Files 1, 2 and 3 respectively. All *S. uvarum* strains are derivatives of the sequenced strain 6 CBS 7001 (previously sometimes called *S. bayanus* or *S. bayanus* var *uvarum*), and all *S.* 7 cerevisiae strains are of S288C background. Unless specified below, yeast strains were 8 grown at 25°C for S. uvarum strains and 30°C for S. cerevisiae strains and standard media 9 recipes were used. 10 11 **Construction of the Tn7 mutagenesis library** 12 The construction of the Tn7 plasmid library has been previously described in detail 13 and was obtained from the Caudy lab (Caudy et al. 2013). Briefly, this mutagenesis 14 approach uses a plasmid library of *S. uvarum* genomic DNA, containing random Tn7

15 transposon insertions. The construct has a selectable marker for transformation into yeast,

16 allowing the selection of disruption alleles.

17 To make the plasmid library, genomic DNA was isolated and fragmented by 18 sonication to an average length of 3 kb from a *rho0 S. uvarum* strain. The ends of the DNA 19 were blunted and cloned into the pZero Blunt vector (Invitrogen). Approximately 50,000 20 colonies were recovered from the transformation into *E. coli* DH5 α strain. The 21 transformants were scraped from Kanamycin plates and pooled for plasmid purification. A 22 version of the Tn7 transposon was constructed by amplifying the promoter from the Tet-on 23 pCM224 (Bellí et al. 1998). The cassette of the Tet-on promoter and the ClonNAT resistance 24 gene was amplified using PCR primers containing lox and BamHI sites and cloned into the 25 BamHI site of the NEB vector pGPS3. This transposon construct was inserted into the S.

1	uvarum genomic DNA library in vitro using the transposon kit from NEB. Initial selection
2	(50,000 colonies) was on ClonNAT/Zeo. HindIII and XbaI were used to digest the pZero
3	backbone to release the linearized genomic DNA for efficient recombination. The library
4	was then used to transform a haploid <i>S. uvarum</i> strain (ACY12) and a diploid strain
5	(YMD1228) using a modified transformation protocol optimized for <i>S. uvarum</i> (Caudy et al.
6	2013). Transformant colonies were plated to YPD-ClonNat plates and allowed to grow for 5
7	days at 25°C. A total of \sim 500,000 colonies were scraped for each pool. Each final pool was
8	well mixed at a 1:1 ratio with 50% glycerol and 2 ml aliquots were stored at -80°C.
9	
10	Pooled growth of Tn7 <i>S. uvarum</i> libraries
11	To determine the initial complexity of the integrated pools, genomic DNA was
12	extracted directly from the glycerol stocks of both haploid and diploid pools using the
13	Hoffman and Winston method (Hoffman and Winston 1987). Additionally, we inoculated
14	$500\ \mu l$ of both libraries in separate YPD flasks for 24 hours to recover mutants after 24
15	hours of growth. Furthermore, to collect samples over time, we competed both pools under
16	sulfate-limiting conditions in chemostats for approximately 30 generations at 25°C. A large-
17	volume, \sim 300ml, sulfate-limited chemostat (Gresham et al. 2008) was inoculated with a
18	single 2ml glycerol stock sample of each pool. After allowing the chemostat to grow at 25°C
19	without dilution for \sim 24 hrs, fresh media was added to the chemostat at a rate of 0.17 h-1.
20	This pooled growth assay was repeated twice, each including 5 time points with O.D. and
21	dilution rate measurements as well as collected cell pellets for DNA extractions using the
22	modified Hoffman-Winston prep referenced above.
23	

Tn7 sequencing library preparation

1	Sequencing libraries were prepared by first extracting genomic DNA from pools of
2	each library grown in YPD and sulfate limited conditions. Genomic DNA libraries were
3	prepared for Illumina sequencing using a Tn7-seq protocol described previously (Wetmore
4	et al. 2015). Briefly, the Covaris was used to randomly fragment DNA to approximately 200-
5	800 bp in length. The fragments were blunt ended and A-tails were added to the fragments
6	to ligate the Illumina adapter sequences. Custom index primers (listed in Supplementary
7	File 3) targeting Tn7-specific sequence and Illumina adapter sequence were used to enrich
8	for genomic DNA with Tn7 insertion sites. The barcoded libraries were quantified on an
9	Invitrogen Qubit Fluorometer and submitted for 150 bp-paired end sequencing on an
10	Illumina HiSeq 2000 by JGI. This method was also applied to make the plasmid library, from
11	linearized plasmid DNA.
12	
13	Sequencing analysis
13 14	Sequencing analysis Sequencing reads from the FASTQ files were trimmed to remove Tn7 specific
13 14 15	Sequencing analysis Sequencing reads from the FASTQ files were trimmed to remove Tn7 specific sequences and adapter sequences, restricting the minimal length of reads to 36 bp using
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13 14 15 16 17 18 19 20 21 22 23 24	Sequencing analysis Sequencing reads from the FASTQ files were trimmed to remove Tn7 specific sequences and adapter sequences, restricting the minimal length of reads to 36 bp using Trimmomatic (Bolger et al. 2014) and FASTX-Toolkit. Trimmed FASTQ files were aligned against the reference strain of <i>S. uvarum</i> (CBS 7001) using Burrows-Wheeler Aligner (BWA) with standard filters applied (Li and Durbin 2009). Specifically, non-uniquely mapping reads, reads in which the pair did not map, reads with a mapping quality less than 30 and PCR/optical duplicate reads were filtered out; the samtools C-50 filter was applied as recommended for reads mapped with BWA. To limit the insertional analysis to actively growing cells, SAM files were merged from the later time points in the growth assays of each pool using samtools (Li et al. 2009). The sequence coverage of the nuclear genome ranged from 70 to 300x (Supplementary File 4). Insertion sites were determined from

processed through a custom Python script that counted the number of insertion events in
 each coding region across the genome. This pipeline was applied to both libraries and
 further comparisons were made between the pools to determine essential genes. Read data
 have been deposited at the NCBI under the SRA accession number SRP115313.

5

6 **Predicting gene dispensability between species**

7 In order to determine a list of predicted essential genes, comparisons were made 8 between the haploid and diploid libraries. We calculated an insertion ratio by dividing the 9 number of insertions in the haploid pool by the number in the diploid pool. This direct 10 comparison inherently accounts for the length of the gene, since the length is constant in 11 both libraries. Therefore, a decrease in insertion sites in the haploid library indicates a 12 reduction in the presence of mutants containing insertional sites that impact cellular 13 viability. Ratios closer to zero represent insertional mutants that reduce the frequency of 14 haploids harboring insertional sites in a coding region that is required for cellular growth. 15 To make an insertion ratio cut-off value to categorize essential and non-essential 16 genes, we analyzed the distribution of insertion ratios within intergenic regions between 17 500 bp and 7 kb in length and positioned between Watson and Crick coding regions (so 18 chosen because these are less likely to contain promoter sequences). The distribution of the 19 insertion ratio calculated for these regions was similar to that of known non-essential genes 20 in *S. cerevisiae*. Therefore, we used this distribution to rank the insertion ratios of all coding 21 regions and set a cut-off value to 0.25 where 20% of the insertion ratio of coding regions fell 22 below the intergenic distribution, which was similar to the kernel density estimates of 23 known S. cerevisiae essential genes. The kernel density estimates were computed in R using 24 X and visualized using ggplot2. To remove a class of low coverage genes in the essential 25 gene category, we applied an additional cut-off value. Since the difference between 0 and 1

with a gene that is longer has a lower weighted difference than a shorter gene, we
 calculated the difference between the diploid pool and haploid pool and normalized this
 value to the length of the gene (normalized difference). Genes with less than a normalized
 difference of 2 were removed from the essential category.

5

6 Validating predicted essential and non-essential genes

7 We validated predicted essential genes by creating *S. uvarum* heterozygous diploid 8 deletion mutants using primers listed in the **Supplemental File 3.** Primers containing 50 9 bp of homology upstream and downstream of each candidate open reading frame were used 10 to amplify the KanMX cassette from the pRS400 plasmid. The PCR product was used to 11 integrate into the *S. uvarum* genome using an *S. uvarum* specific transformation protocol. 12 The proper integration of the construct was validated through clone purifying positive 13 colonies and extracting genomic DNA to perform PCR using diagnostic primers listed in the 14 **Supplemental File 3.** The diagnostic primers were designed to target ~ 150 bp upstream 15 and \sim 150 bp downstream of the open reading frame to identify wild type and drug marker 16 alleles. Positive clones were sporulated for 3-5 days at 25°C and 8 tetrads were screened 17 for 2:2 viable segregation. Images were taken after 4 days of growth on YPD plates. 18 Mutants conferring non-essential phenotypes were replicated on G418 plates and images 19 were taken after 4 days of growth at 25°C (**Supplemental Fig. 9&10**). This method was 20 also applied to making double mutants. A collection of 440 MAT α S. uvarum strains was 21 generated by standard methods in the Rine lab and used as confirmed non-essential genes. 22

23 **Cross-species complementation assays**

To determine if genes are diverging in gene function or in other trans-acting factors,
we performed cross-species complementation assays with species-specific essential genes.

1	Essential genes that were <i>S. cerevisiae</i> specific were tested in a heterozygous diploid
2	deletion strain from the magic marker collection. Alleles of each S. cerevisiae essential gene
3	were amplified from <i>S. cerevisiae</i> and <i>S. uvarum</i> genomes and cloned into a CEN ARS
4	plasmid. PCR using Phusion DNA polymerase was used to amplify 500 bp upstream and 5
5	bp downstream of the stop codon of each gene from <i>S. cerevisiae</i> and <i>S. uvarum</i> . Each gene
6	was cloned into pIL37 by Gibson assembly using primers listed in Supplemental File 3
7	using standard methods (Thomas et al. 2015). All plasmids used in this study are listed in
8	Supplemental File 2. The S. cerevisiae heterozygous diploid deletion strains were
9	transformed with a plasmid containing a corresponding allele from each species and
10	selected on C-URA plates. Similarly, S. uvarum specific essential genes were also tested by
11	making each heterozygous diploid deletion strain $ura3\Delta/ura3\Delta$, and transformed with a
12	plasmid containing a corresponding <i>S. cerevisiae</i> allele from the MoBY-ORF collection (Ho et
13	al. 2009).
14	Transformed strains were sporulated for 5 days at 30°C and 25°C for <i>S. cerevisiae</i> and
15	S. uvarum species, respectively and tetrad analysis was performed on YPD plates. After 3
16	days of growth, plates were replica plated on C-URA and YPD+G418 plates and imaged after
17	2 days of growth (Supplemental Fig. 15).
18	
19	Data Access
20	The transposon insertion data in this study have been submitted to the NCBI Sequence Read
21	Archive (SRA; https://www.ncbi.nlm.nih.gov/sra/) under SRA accession number
22	SRP115313.

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17	
10	

- 18 **Competing financial interests**
- 19 The authors declare no competing financial interests.
- 20

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