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A new method for determining ribosomal DNA copy number shows differences between *Saccharomyces cerevisiae* populations

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Abstract

Ribosomal DNA genes (rDNA) encode the major ribosomal RNAs (rRNA) and in eukaryotic genomes are typically present as one or more arrays of tandem repeats. Species have characteristic rDNA copy numbers, ranging from tens to thousands of copies, with the number thought to be redundant for rRNA production. However, the tandem rDNA repeats are prone to recombination-mediated changes in copy number, resulting in substantial intra-species copy number variation. There is growing evidence that these copy number differences can have phenotypic consequences. However, we lack a comprehensive understanding of what determines rDNA copy number, how it evolves, and what the consequences are, in part because of difficulties in quantifying copy number. Here, we developed a genomic sequence read approach that estimates rDNA copy number from the modal coverage of the rDNA and whole genome to help overcome limitations in quantifying copy number with existing mean coverage-based approaches. We validated our method using strains of the yeast *Saccharomyces cerevisiae* with previously-determined rDNA copy numbers, and then applied our pipeline to investigate rDNA copy number in a global sample of 788 yeast isolates. We found that wild yeast have a mean copy number of 92, consistent with what is reported for other fungi but much lower than in laboratory strains. We show that different populations have different rDNA copy numbers. These differences can partially be explained by phylogeny, but other factors such as environment are also likely to contribute to population differences in copy number. Our results demonstrate the utility of the modal coverage method, and highlight the high level of rDNA copy number variation within and between populations.

Introduction

The ribosomal RNA gene repeats (rDNA) encode the major ribosomal RNA (rRNA) components of the ribosome, and thus are essential for ribosome biogenesis and protein translation. In most eukaryotes the rDNA forms large tandem repeat arrays on one or more chromosomes [1]. Each repeat unit comprises a coding region transcribed by RNA polymerase I (Pol-I) that encodes 18S, 5.8S and 28S rRNA [2], and an intergenic spacer region (IGS) that separates adjacent coding regions (**Fig 1**). The number of rDNA repeat copies varies widely between species, typically from tens to hundreds of thousands of copies [1, 3-5]. However, each species appears to have a 'set' or homeostatic (in the sense of [6]) rDNA copy number that is returned to if the number of copies deviates [7-10]. Deviation in rDNA copy number between individuals within a species is well documented and can be substantial [11-16]. This copy number variation is thought to be tolerated because of redundancy in rDNA copies [8, 17]. This redundancy can partly be explained the striking observation that only a subset of the repeats is transcribed at any one time [2]. Thus, cells can compensate for changes in rDNA copy number by activating or silencing repeats to maintain the same transcriptional output [18]. Variation in rDNA copy number is a consequence of unequal homologous recombination, which results in loss or gain of rDNA copies [8, 19-22]. This copy number variation is, somewhat counter-intuitively, what drives the high levels of sequence homogeneity observed between the rDNA copies within a genome, a pattern known as concerted evolution [23-25]. Recent results in *Saccharomyces cerevisiae* revealed an elegant mechanism through which homeostatic rDNA copy number is maintained in the face of rDNA copy number change via the abundance of the Pol-I transcription factor UAF (upstream activating factor) and the histone deacetylase Sir2 [26]. However, the selective pressure(s) that determines what the homeostatic rDNA copy

71 number is remains unknown. Nevertheless, there is growing evidence that rDNA copy

72 number and the proportion of active/silent rDNA copies impact several aspects of cell

73 biology beyond simply rRNA production [8, 12, 17, 22, 27-35].

Figure 1. Organization of the rDNA repeats in *Saccharomyces cerevisiae*. Top shows

a schematic of tandemly-repeated units in the rDNA array located on chromosome XII.

Bottom shows the organization of an individual rDNA repeat including transcription start

sites, the 5' external transcribed spacer (5'ETS), the rRNA (18S, 5.8S and 28S) coding

genes, the two internal transcribed spacers (ITS1 and 2), and the intergenic spacer (IGS).

The IGS is divided into two by a 5S rRNA gene. Schematic is not to scale.

Interest in the phenotypic consequences of rDNA copy number variation has led to a

number of approaches being used to measure it. These include molecular biology

approaches such as quantitative DNA hybridization [36-39], pulsed field gel

electrophoresis (PFGE) [40, 41], quantitative real-time PCR (qPCR) [15, 42-46] and digital

droplet PCR (ddPCR) [47, 48]. A major advance in the measurement of rDNA copy number

has been the emergence of bioinformatic approaches that use whole genome (WG) next

generation sequencing (NGS) reads to estimate copy number, based on the rationale that

sequence coverage of the rDNA correlates with copy number. This correlation is a

consequence of concerted evolution, with the high sequence identity between repeats

resulting in reads from all rDNA copies mapping to a single reference rDNA unit, thus

providing a high coverage signal that is proportional to copy number. Existing

bioinformatic approaches calculate the mean rDNA read coverage and normalize to the

mean WG coverage to estimate copy number [5, 12, 25, 34, 49], thus assuming that mean

coverage represents the "true coverage" for both the rDNA and the WG. However, there are

reasons to suspect this mean coverage approach assumption might not always hold.

Repetitive elements (e.g. microsatellites and transposons), PCR/sequencing bias (which is

particularly evident for the rDNA [50-54]; **Supplementary Figure 1**), and large-scale

mutations such as aneuploidies and segmental duplications may all cause the measured

mean coverage to differ from the real coverage. While efforts have been made to address

some of these potential confounders [12, 55, 56], estimated copy number varies depending

on which region of the rDNA is used [12, 34], thus the accuracy of this mean read coverage

approach has been called into question [5, 46].

Here we present a bioinformatics pipeline that measures rDNA copy number using modal

(most frequent) NGS read coverage as a way to overcome the limitations of the mean

coverage bioinformatics approach. We assessed the parameters important for performance

and validated the pipeline using *S. cerevisiae* strains with known rDNA copy numbers. We

then employed our pipeline to investigate whether *S. cerevisiae* populations maintain

different homeostatic rDNA copy numbers.

Materials and Methods

Modified *Saccharomyces cerevisiae* genome

Chromosome sequences for W303 were obtained from the NCBI (accession CM001806.1 - CM001823.1) and concatenated. rDNA copies present within the W303 reference genome were identified using BLAST and removed using Geneious (v. 11.0.3). The *S. cerevisiae* W303 strain rDNA repeat unit from [23] was added as an extrachromosomal rDNA reference, and this modified W303 yeast reference genome (W303-rDNA) was used in subsequent analyses.

Yeast strains/isolates and growth conditions

Yeast strains/isolates that were cultured are listed in **Table 1**. Culturing was performed in liquid or solid (2% agar) YPD (1% w/v yeast extract, 2% w/v peptone and 2 % w/v D+ glucose) medium at 30°C.

Table 1: *S. cerevisiae* strains/isolates cultured in this study

Strain/isolate	Details	Source
Wild-type	*MATa ade2-1 ura3-1 his3-11 trp1-1 leu2-3, 112 can1-100 fob1Δ::HIS3*	NOY408-1bf; [17]
20-copy	*MATa ade2-1 ura3-1 his3-11 trp1-1 leu2-3, 112 can1-100 fob1Δ::HIS3*	[17]

40-copy	*MATa ade2-1 ura3-1 his3-11 trp1-1 leu2-3, 112 can1-100 fob1Δ::HIS3*	[17]
80-copy	*MATa ade2-1 ura3-1 his3-11 trp1-1 leu2-3, 112 can1-100 fob1Δ::HIS3*	[17]
YJM981	Human clinical isolate from Italy; *Mat **a**, ho::HygMX, ura3::KanMX*-Barcode	[57]
DBVPG1373	Netherlands isolate from soil; *Mat **a**, ho::HygMX, ura3::KanMX*-Barcode	[57]
UWOPS03-461-4	Malaysian isolate from nectar	[58]
UWOPS03-461-4 (Mat a)	Derivative of UWOPS03-461-4; *Mat **a**, ho::HygMX, ura3::KanMX*-Barcode	[57]
UWOPS03-461-4 (Mat α)	Derivative of UWOPS03-461-4; *Mat α, ho::HygMX, ura3::KanMX*-Barcode	[57]
YPS128	US isolate from soil beneath *Quercus alba*	[58]
DBVPG1788	Finland isolate from vineyard soil	[58]

Genomic DNA extraction

High molecular weight genomic DNA (gDNA) was isolated as follows. Cell pellets from 3–5

mL liquid cultures were washed in 500 µL of 50 mM EDTA pH 8 and resuspended in 200 µL

of 50 mM EDTA pH 8 supplemented with zymolyase (3 mg/mL). After 1 hr at 37°C, the cell

lysate was mixed with 20 µL of 10% sodium dodecyl sulfate then with 150 µL of 3 M

potassium acetate (KAc) and incubated on ice for 1 hr. 100 µL of phenol-chloroform-

isoamyl alcohol was added to the SDS-KAc suspension, and following vortexing and

centrifugation, 600 µL of propanol-2 were added to the aqueous supernatant ($\approx$ 300 µL).

The nucleic acid pellet was washed three times in 70% EtOH, dried and resuspended in

PCR grade water supplemented with RNase A (0.3 mg/mL). After 1 hr at 37°C, samples

were stored at -20°C.

Whole genome sequence data

gDNA extracted from four isogenic strains with different rDNA copy numbers (WT, 20-

copy, 40-copy and 80-copy; **Table 1**) was sequenced using Illumina MiSeq

(**Supplementary Table 1**). The raw sequence files are available through the NCBI SRA

(accession number SUB7882611).

Read preparation

Paired-end reads were combined and quality checked using SolexaQA [59]. Low-quality

ends of reads (score cutoff 13) were trimmed using DynamicTrim, and short reads were

removed using a length cutoff of 50 bp with LengthSort, both within SolexaQA, as follows:

command: ~/path/to/solexaQA/SolexaQA++ dynamictrim /fastq/file

command: ~/path/to/solexaQA/SolexaQA++ lengthsort -l 50 /trimmed/fastq/file

Obtaining whole genome and rDNA coverages

The W303-rDNA reference genome was indexed using the bowtie2 (v. 2.3.2) build

command as follows:

command: ~/bowtie2-2.3.2/bowtie2-build <reference_in> <bt2_base>

Coverage files for the whole genome and rDNA were obtained using a four step pipeline:

Step-1: Processed reads were mapped to the indexed W303-rDNA genome using bowtie2

(v. 2.3.2)

command:　　　~/bowtie2-2.3.2/bowtie2　　-x　　　/path/to/indexed/genome/　　-U

/path/to/trimmed/reads/ -S /output SAM file/

Step-2: The subsequent SAM format alignment was converted to BAM format using the

SAMtools (v. 1.8) view command:

command: ~/samtools-1.8/samtools view -b -S -o <output_BAM> <input_SAM>

Step-3: Mapped reads in the BAM file were sorted according to the location they mapped to

in W303-rDNA using the SAMtools sort command:

command: ~/samtools-1.8/samtools sort <input_BAM> -o <output_sorted.bam>

Step-4: Per-base read coverages across the entire W303-rDNA genome and the rDNA were

obtained using BEDtools (v. 2.26.0):

command:　　　~/bedtools　　　genomecov　　　-ibam　　　<aligned_sorted.bam>　　　-g

<reference_genome.fasta> -d <bedtools_coverage_WG.txt>

command: grep "rDNA_BLAST" <bedtools_coverage_WG.txt>

<rDNA_bedtools_coverage.txt>

Calculation of rDNA copy number using modal coverage

Coverage frequency tables for the rDNA and whole genome (excluding mitochondrial DNA and plasmids) were obtained from per-base read coverage files by computing the mean coverage over a given sliding window size with a slide of 1 bp. The mean coverage for each sliding window was then allocated into a coverage bin. The bin that includes read coverage of zero was removed. The three highest frequency coverage bins from both the rDNA and whole genome frequency tables were used to calculate rDNA copy number as follows:

$$rDNA\ copy\ number\ = \frac{Peak\ rDNA\ coverage\ bin\ value}{Peak\ whole\ genome\ coverage\ bin\ value}$$

The rDNA copy number estimates were taken as the mean of all pairwise combinations of these copy number values (**Supplementary Figure 2**).

Pipeline availability

The pipeline for modal calculation of rDNA copy number from an alignment of sequence reads to a reference genome containing one rDNA copy is available through Github (https://github.com/diksha1621/rDNA-copy-number-pipeline).

Calculation of rDNA copy number using mean and median coverage

The per-base read coverage across W303-rDNA from Bedtools was input into custom R-scripts to obtain the mean and median coverage values for the whole genome and rDNA, after removing the rDNA, 2-micron plasmid, and mitochondrial DNA coverage values from

195   the whole genome calculation. The rDNA copy number was then calculated for both mean

196   and median approaches as follows:

$$rDNA\ copy\ number = \frac{coverage\ across\ rDNA}{coverage\ across\ whole\ genome}$$

198   Subsampling

199   To generate different coverage levels for copy number estimation, sequence reads were

200   randomly downsampled using the seqtk tool (https://github.com/lh3/seqtk) as follows:

201   **command:** ~/seqtk/seqtk sample –s$RANDOM <name of fastqfile> <number of reads

202   required> <outputfile>

204   rDNA copy number measurement by ddPCR

At least three independent cultures (biological replicates) were generated for each isolate

using one independent colony per culture. To evaluate rDNA copy number variation over

generations, cultures were propagated over four days (~60 generations) as follows:

individual colonies were initially grown in 3 mL YPD for 24 hr. 30 μL of this was used to

inoculate 3 mL YPD and this was grown for another 24 hr. This process was repeated for

four days. Cells were harvested after 24 hr (~15 generations) and four days, and cell pellets

frozen at -80˚C. gDNA was extracted as above, then linearized by *Xba*I in NEB2 buffer

following the manufacturer's instructions (NEB) to individualize rDNA repeats. gDNA

linearization was verified by separation on agarose gels and DNA concentration measured

on a Qubit Fluorometer using the Qubit dsDNA HS assay (Thermo Fisher). Linearized gDNA

was brought to 2 pg/µL by serial dilution. EvaGreen master mixes were prepared with an

rDNA primer pair (rDNAScSp_F2 5'- ATCTCTTGGTTCTCGCATCG-3', rDNAScSp_R2 5'-

GGAAATGACGCTCAAACAGG-3') or a single copy *RPS3* gene primer pair (RPS3ScSp_F2 5'-

CACTCCAACCAAGACCGAAG-3', RPS3ScSp_R2 5'-GACAAACCACGGTCTTGAAC-3'). *RPS3* and

rDNA ddPCR reactions were performed with 2 µL (4 pg) of the same linearized gDNA

dilution as template. Droplet generation and endpoint PCR were performed following the

manufacturer's instructions, and droplets were read using a QX200 droplet reader

(BioRad). Quantification was performed using QuantaSoft Analysis Pro (v. 1.0.596). rDNA

copy number was determined by the (rDNA copy/µL)/(*RPS3* copy/µL) ratio.

Pulse field gel electrophoresis (PFGE)

To make chromosome plugs [21], cells from overnight liquid cultures were resuspended in

50 mM EDTA pH 8.0 to 2.10^9 cells/mL, transferred to 45°C, and mixed with an equal

volume of 1.5% low melting point agarose in 50 mM EDTA prewarmed to 45°C. The

mixture was transferred by gentle pipetting to PFGE plug molds (BioRad) to set at 4°C for

15 min. Plugs were transferred to fresh spheroplasting solution (1 M Sorbitol, 20 mM EDTA

pH 8.0, 10 mM Tris-HCl pH 7.5, 14 mM 2-mercaptoethanol, 2 mg/mL zymolyase). After 6 hr

incubation at 37°C with occasional inversion, plugs were washed for 15 min in LDS buffer

(1% lithium dodecyl sulphate, 100 mM EDTA pH 8.0, 10 mM Tris-HCl pH 8.0), before

overnight incubation at 37°C in the same buffer with gentle shaking. Plugs were incubated

twice for 30 min each in NDS buffer (500 mM EDTA, 10 mM Tris-HCl, 1% sarkosyl, pH 9.5)

and at least three times for 30 min in TE (10 mM Tris-HCl pH 8.0, 1 mM EDTA pH 8.0).

Plugs were stored at 4°C in fresh TE. For restriction digestion, half plugs were pre-washed

for two hours in TE, three times for 20 min each in TE, and three times for 20 min each in

300 μL restriction buffer supplemented with 100 μg/mL BSA, all at room temp. Restriction

digestion was performed overnight at the recommended temperature in 500 μL of

restriction buffer containing 100 U of restriction endonuclease. Digested plugs were

washed in 50 mM EDTA pH 8.0 and stored at 4°C in 50 mM EDTA pH 8.0 before loading.

PFGE was performed using 1% agarose gel in 0.5X TBE (Thermo-Fisher) in a CHEF Master

XA 170-3670 system (BioRad) with the following parameters: auto algorithm separation

range 5 kb - 2 Mb (angle 120°C, run 6 V/cm, initial switch time 0.22 s, final switch time 3

min 24 s, run time 916 min) at 14°C. DNA was visualized by staining in ethidium bromide

(5 μg/mL) and imaging (Gel Doc XR+; BioRad).

1002 Yeast Genome project rDNA copy number estimation

Illumina reads from the 1002 Yeast Genome project were obtained from the European

Nucleotide Archive (www.ebi.ac.uk/) under accession number ERP014555. We omitted

clades with few members, mosaic clades, and unclustered isolates, giving a total of 788

isolates. Reads were downsampled to 10-fold-coverage using seqtk() and rDNA copy

number for each isolate was calculated using W303-rDNA as the reference. Bin sizes of

1/200th of the mean coverage for rDNA and 1/50th for the whole genome, and a window

size of 600 bp for both estimates, were used. Violin plots were plotted using the ggplot()

package in R.

Phylogenetic analyses

To create a neighbour-joining phylogeny based on rDNA copy number values, rDNA copy

number for each isolate (after removing 30 isolates for which SNP data were not available)

was normalized on a 0-1 scale. Normalized values were used to calculate pairwise

Euclidean distances between each pair of isolates to generate a distance matrix that was

applied to construct a phylogeny via neighbour-joining using MEGA X [60].

Phylocorrelograms of copy number and the SNP phylogeny were generated using

phylosignal v.1.3 [61] (https://cran.r-project.org/web/packages/phylosignal/index.html).

Phylocorrelograms representing a no phylogenetic signal dataset (a "white noise" random

distribution) and a high phylogenetic signal dataset (a character evolving on the SNP tree

according to a Brownian motion model) were also generated. For the white noise

distribution, data were simulated from a normal distribution with mean and standard

deviation matching those of the observed copy number data (mean=92.5, sd=30.8). For the

Brownian motion model, we first estimated the ancestral mean ($z0=83.2$) and the rate

parameter ($\sigma2=72557.2$) from the observed copy number data using the fitContinuous

function from geiger [62] (https://cran.r-project.org/package=geiger). Then, we simulated

from these parameters on the SNP tree using fastBM from phytools 0.7 (https://cran.r-

project.org/package=phytools). Phylocorrelograms were generated for the observed and

the two simulated datasets, estimating correlations at a series of 100 phylogenetic

distances using 100 bootstrap replicates.

Comparing intra-species variation in rDNA copy number

Copy number estimates for twelve isolates from the 1002 Yeast Genome data were

randomly drawn 1000 times using a custom bash-script to obtain rDNA copy number

ranges.

Statistical analyses

All statistical analyses to evaluate differences in rDNA copy number between clades were

performed in R. Significance was calculated using the Welch t-test (t-test), the non-

parametric Wilcoxon-Mann-Whitney test (wilcox test) and ANOVA, with p-values

considered statistically significant at $p < 0.05$.

Results and Discussion

Establishment of a modal coverage bioinformatics pipeline for estimating rDNA copy

number

The abundance of data generated from NGS platforms has led a number of studies to use

mean read depth to estimate rDNA copy number [5, 12, 25, 34, 49, 55, 56]. However, repeat

elements, sequence biases and large-scale changes like aneuploidies can potentially result

in non-normal read coverage distributions where the mean coverage does not accurately

represent the true coverage. To overcome these limitations, we developed a novel

sequence read-based rDNA copy number calculation approach based on the most frequent

(modal) coverage. The rationale for this approach is that modal coverage will provide an

estimate of the relative coverage representation of a given region in a genome that is more

robust to biases away from normality than the mean or median, The approach allocates

coverage across a reference genome into coverage bins. The ratio of the most frequently

occurring coverage bins for the rDNA and the WG is then used to calculate rDNA copy

number (per haploid genome). We implemented this modal coverage approach as a simple

pipeline to calculate rDNA copy number from mapped sequence reads (**Fig 2**). To help

smooth across positions that stochastically vary in coverage, an issue that is particularly

prevalent with very low coverage datasets, we used a sliding window approach to calculate

coverage. Our straightforward pipeline uses a sorted BAM file of reads aligned to a

reference genome for which the position of the rDNA is known (either embedded in the

genome or as a separate contig) to calculate copy number

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Figure 2. Overview of the modal approach to estimate rDNA copy number from whole

genome sequence data. Whole genome (WG) sequence reads are mapped against a

reference genome containing a single rDNA copy. Mean read depth for each postion is

calculated across the rDNA and the WG using a sliding window, then allocated into

coverage bins (shown as histograms). To calculate modal rDNA copy number, the highest

frequency coverage bins for both the rDNA and WG are used to compute ratios that

represent the rDNA copy number range. The histograms shown were plotted using a 20-

copy yeast strain at 5-fold WG coverage with bin sizes of $1/200^{th}$ of mean coverage for

rDNA and $1/50^{th}$ for WG, and a sliding window of 600 bp for both. The coverage ranges for

the three most frequent bins for each are indicated in boxes.

To implement our modal coverage approach, we generated test datasets by performing WG

Illumina sequencing of a haploid wild-type laboratory *S. cerevisiae* strain reported to have

150-200 rDNA copies, and three isogenic derivatives where the rDNA has been artificially

reduced to 20, 40 and 80 copies, and "frozen" in place through disruption of a gene (*FOB1*)

that promotes rDNA copy number change [17] (**Table 1**). Initially, we investigated which

parameters provide the most accurate results by applying our pipeline to the WG sequence

data obtained from a strain with 20 rDNA copies (20-copy strain). We obtained a genome-

wide read coverage of 13.1-fold (**Supplementary Table 1**) and mapped these reads to the

W303-rDNA yeast reference genome that has a single rDNA copy. The mapping output was

used to determine per-base coverage values, which were placed into coverage bins using a

sliding window. We investigated a range of sliding window sizes, from 100 bp (previously

reported to have an approximately normal distribution of WG sequence read coverage

[63]) to 1,000 bp (large sliding window sizes, whilst smoothing stochastic coverage

variation, converge on the mean coverage as the window size approaches the rDNA unit

length). We also assessed the impact of coverage on copy number estimation by

downsampling the sequence reads. We ran analyses with 100 technical replicates and

computed the rDNA copy number means and ranges. We found that, as expected, the

accuracy and precision (defined here as <u>similarity to known copy number</u> and <u>copy</u>

<u>number range</u>, respectively) of the pipeline was poorer at lower coverage levels, while

larger sliding window sizes could compensate for a lack of reads to improve both measures

(**Fig 3**). Coverage levels above 10-fold with a sliding window size between 500-800 bp

produced accurate rDNA estimates. However, our method also demonstrated adequate

performance even with a coverage level of 5-fold, when the sliding window was 600-700

bp (**Fig 3**). We found that the method works similarly when just using the rRNA coding

region (**Supplementary Figure 3**) rather than the full repeat, which is important as the full

rDNA unit sequence is often not available. We also examined the performance of median

coverage, but found that while it had greater precision compared to the modal coverage

approach, the accuracy was poorer (**Supplementary Figure 4**). Given the rapid rate at

which copy number changes even during vegetative growth [21], the lower precision of our

method may more accurately represent the range of copy numbers likely to be present in

samples that consist of multiple cells.

**Figure 3. Assessing parameters for rDNA copy number estimation accuracy and
precision.** Cells represent the (**A**) deviation of the calculated modal rDNA copy number
from 20, and (**B**) maximum variation of rDNA copy number calculated from the 100
technical replicates for each coverage level and sliding window (SW) size combination. The
heatmap scales used are indicated. In (**A**), rDNA copy number was rounded to the nearest
integer.

We then assessed the performance of our pipeline with the 40-copy, 80-copy, and WT *S.*

cerevisiae strain data. Illumina WG sequence reads (**Supplementary Table 1**) obtained

from these strains were downsampled to generate 100 technical replicates at 10-fold

coverage for each strain, and rDNA copy numbers were calculated using our modal

coverage pipeline with a sliding window of 600 bp. The resultant rDNA copy numbers

were: 32-40 ($\bar{x}$ = 36 copies) for the 40-copy strain; 57-72 ($\bar{x}$ = 64 copies) for the 80-

copy strain; 129-177 ($\bar{x}$ = 157 copies) for the WT strain. These values, while similar to the

reported copy numbers for these strains, are not identical. Therefore, to check the actual

copy numbers of these strains, and to provide a direct validation of our modal pipeline

method, we next experimentally determined the rDNA copy numbers of these strains.

We chose ddPCR to experimentally determine rDNA copy number because it is less

sensitive than qPCR to biases in secondary structure regions that are common in the rDNA

coding region [22]. The ddPCR data showed that the rDNA copy numbers of our strains are

similar to those calculated by our modal coverage method, with both methods suggesting

that the "80-copy" strain actually has substantially fewer copies than reported (**Table 2**;

Supplementary Figure 5A), perhaps due to a stochastic change in copy number that has

occurred in our version of this strain. We also compared our modal coverage approach

with the mean coverage calculated from the same datasets. We used a simple mean

calculation to match the implementation of our modal approach, using the same down-

sampled 10-fold WG coverage datasets. The copy number estimates made using the mean

coverage approach were uniformly lower than the other estimates (**Supplementary Table**

2), which we suggest is the result of sequencing biases against regions in the rRNA coding

region. Importantly, correlating read coverage and ddPCR copy number estimates showed

the modal coverage slope was closer to the expected value of 1 than the mean coverage

slope (**Fig 4**). We also estimated the copy number using pulsed field gel electrophoresis

based on the size of the restriction fragment encompassing the entire rDNA array divided

by the rDNA unit size (accounting for the sizes of the flanking regions), again with

consistent results (**Supplementary Figure 5B,C**). Together, these results suggest the

modal coverage approach is an accurate way to estimate rDNA copy number.

Figure 4. Comparison of modal and mean coverage copy number estimation methods. Plot of rDNA copy number for the 20, 40, 80 and WT *S. cerevisiae* strains (10-fold coverage) calculated using modal (orange line) and mean (blue line) coverage methods versus the copy number determined by ddPCR. The expected 1:1 correlation between read coverage and ddPCR methods is shown in black. Note that while the mean coverage method gives a higher R^2, the modal coverage results are a closer fit to the expected 1:1 line.

Our results suggest that the modal coverage pipeline provides robust estimates of rDNA copy number even when coverage is less than 5-fold. This reliability may partly be a consequence of the larger sliding window size we used compared to that commonly applied for mean coverage methods. It was previously reported that coverage below ~65X

results in precision issues when estimating rDNA copy number [5]. However, we did not

find this, either for our method or using mean coverage, suggesting that the issues might be

specific to the approach or dataset used in that study. The simple implementation of our

modal approach coupled with its good performance make it an attractive method for

estimating rDNA copy number from sequence read data. Furthermore, a modal approach is

expected to be more robust to features that can perturb mean coverage approaches by

skewing coverage distributions, such as repeat elements, large duplications and deletions,

regions exhibiting sequencing biases, modest sequence divergence from the reference

sequence, and aneuploidies [46]. Although we have developed our pipeline for measuring

rDNA copy number, in principle it can be used to calculate copy number for any repeated

sequence where all reads map to a single repeat copy and the sequence is known, such as

mitochondrial and chloroplast genome copy numbers. Given its strong performance, we

applied our method to characterize the inter-population distributions of rDNA copy

number in *S. cerevisiae.*

Within-species evolutionary dynamics of rDNA copy number

Studies in model organisms have provided evidence that each species has a homeostatic

copy number which is returned to following copy number perturbations [7-10]. This

homeostatic copy number appears to have a genetic basis [5, 26], which suggests it might

vary between populations, as well as between species. However, few studies have

addressed this question. Given that variation in rDNA copy number has been associated

with altered phenotypes [8, 12, 17, 22, 27-35], we decided to undertake a comprehensive

assessment of *S. cerevisiae* rDNA copy number at the population level using the global wild

yeast dataset from the 1002 Yeast Genome project [64].

We obtained WG sequence data for 788 isolates from the 1002 Yeast Genome project.

Reads for each isolate were downsampled to 10X genome coverage, mapped to our W303-

rDNA reference genome, and rDNA copy numbers estimated using our modal coverage

pipeline. The rDNA copy numbers ranged between 22-227 ($x̄$ = 92) across the 788

isolates (**Supplementary Table 3**). The copy numbers of 11 wild *S. cerevisiae* isolates

included in our dataset had previously been estimated [14, 25], and our results are largely

consistent with these (**Supplementary Table 4**). However, the copy number we estimate

are, in general, much lower than those (~150-200) measured for most laboratory strains

(e.g. [17, 38, 41]). We looked to see whether ploidy affects rDNA copy number, given that

laboratory strains are predominantly haploid, while the wild *S. cerevisiae* isolates we

analyzed are mostly diploid. We observed a small difference in copy number between

haploid and diploid isolates (104 vs 91 copies, respectively; **Supplementary Figure 6** and

Supplementary Information), but overall do not find a strong effect of ploidy on copy

number. Thus, the copy number differences between lab and most wild *S. cerevisiae*

isolates seem to be a property of these isolates.

The difference in copy number between lab and wild *S. cerevisiae* isolates suggests that *S.*

cerevisiae populations may harbor different rDNA copy numbers. To test this, we used the

23 phylogenetic clades defined by [64] as proxies for *S. cerevisiae* populations and looked

at the distributions of rDNA copy number number within and between these populations

(**Fig 5**). ANOVA analysis rejects homogeneity of rDNA copy number between these

populations (p = 4.37e[-15]), suggesting there are population-level differences in copy

number within *S. cerevisiae*.

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Fig 5. rDNA copy number in *S. cerevisiae* populations. To the left is the phylogeny of

the 23 *S. cerevisiae* clades from [64] that encompass the 788 isolates included in this

study. The scale represents substitutions per site. On the right, rDNA copy number

calculated using the modal coverage method is displayed as a violin plot for each clade with

mean population copy numbers indicated by white triangles. Numbers to the right represent

the number of isolates in each clade. The red vertical line represents the overall mean

rDNA copy number (92 copies). Copy number estimations were determined using 10-fold

coverage and a 600 bp sliding window.

We next wanted to look for complementary evidence that *S. cerevisiae* populations have

different rDNA copy numbers, as an alternative explanation for our results is different

populations happened to have different copy numbers simply due to stochastic variation

[21]. If the stochastic variation explanation is correct, we would expect divergent copy

numbers to return to the homeostatic value over time. To test this, we used ddPCR to

measure the rDNA copy numbers of six of the 1002 Yeast Genome project isolates that

represent the range of copy numbers observed, including one that we had three different

isolates of. We grew three biological replicates of each isolate for ~60 generations to allow

any fluctuation in rDNA copy number to return to the homeostatic level [7]. The rDNA copy

numbers, both before and after the ~60 generations, resemble the copy numbers we

estimated from the sequence data and show no tendency to converge on the overall *S.*

cerevisiae mean copy number (**Table 3; Supplementary Table 5**). These results strongly

suggest that our method of estimating rDNA copy number is robust and that the copy

numbers of isolates are not recovering towards a common copy number value. From this

we conclude that different *S. cerevisiae* populations have different homeostatic rDNA copy

numbers.

Table 3. *S. cerevisiae* **rDNA copy number does not recover to a common value**

following ~60 generations of growth

Isolates	rDNA CN at start[a]	rDNA CN after ~60 generations[a]		Original modal CN estimation[b]
S. cerevisiae **wild-type** rep1[c]	213	130	185[d]	157
rep2		217		
rep3		208		
YJM981 rep1	174	120	175	171
rep2		183		
rep3		221		
DBVPG1373 rep1	69	77	85	78
rep2		72		
rep3		107		
UWOPS03-461-4[e] rep1	85	113	95	106
rep2		88		
rep3		83		
UWOPS03-461-4[e] **(Mata)** rep1	244	164	146	
rep2		167		
rep3		106		
UWOPS03-461-4[e] **(Matα)** rep1	ND[f]	108	109	
rep2		115		
rep3		105		
YPS128 rep1	89	87	79	89
rep2		73		
rep3		77		
DBVPG1788 rep1	95	126	108	87
rep2		100		
rep3		97		

[a] Measured using ddPCR

[b] Measured using our modal coverage pipeline

[c] rep: biological replicate

[d] Mean of the three replicates to the nearest integer

[e] UWOPS03-461-4 is the parent isolate of UWOPS03-461-4 Mat**a** and UWOPS03-461-4
Matα derivatives

[f] Not determined

Copy number has previously been shown to correlate with phylogeny for species across the

fungal kingdom [5]. Given the differences in rDNA copy number we observe, we wondered

whether a similar correlation exists between *S. cerevisiae* populations. To test this, we

constructed a neighbour-joining phylogeny using rDNA copy number as the phylogenetic

character for 758 isolates (30 were removed as SNP data were not available) and

compared this to the reported *S. cerevisiae* phylogeny created from genomic SNP data [64].

To assess how well the two phylogenies correlate, we used Moran's Index of spatial

autocorrelation *I*, which quantifies the correlation between two traits. Moran's *I* indicated a

modest positive correlation between rDNA copy number and phylogeny at short

phylogenetic distances (**Fig 6**), but not a significant negative correlation at greater

phylogenetic distances like that previously observed above the species level [5]. These

results suggest that phylogeny only partially explains the distribution of rDNA copy

numbers amongst *S. cerevisiae* populations.

Figure 6. **Phylocorrelograms of autocorrelation based on Moran's *I*.** Phylogenetic

distance spatial autocorrelations between the SNP-based *S. cerevisiae* phylogeny and the

rDNA copy number phylogeny (**A**), a Brownian motion phylogeny (**B**), and random data (**C**)

are plotted. Red segments beneath each phylocorrelogram indicate significant positive

autocorrelation; black no significant autocorrelation, and blue significant negative

autocorrelation. Dotted lines indicate autocorrelation 95% confidence intervals. Significance

is based on comparison to zero phylogenetic autocorrelation (horizontal black line at 0).

Another feature that might explain the distribution of rDNA copy numbers between *S.*

cerevisiae populations is the environment, given that nutritional conditions have been

proposed to influence copy number [65, 66]. To investigate this, we compared the rDNA

copy numbers from two phylogenetically-diverged *S. cerevisiae* populations that are

associated with oak trees, which we took as a proxy for similar environments. We found

the oak populations did not show significantly different copy numbers (p-value = 0.52), as

expected if environment is contributing to copy number. Thus, rDNA copy number might

be partially determined by the environmental conditions the population has evolved in.

However, we found no consistent pattern of similarities or differences with the copy

numbers of the nearest phylogenetic neighbours of these oak clades (**Supplementary**

Information), thus we cannot rule out these results simply representing stochastic

variation. We suggest that a better understanding of what environmental factors modulate

rDNA copy number is necessary before we can properly evaluate the impact of the

environment on patterns of rDNA copy number variation.

Finally, we wondered whether large range in estimated *S. cerevisiae* rDNA copy number

(22-227 copies) might reflect an unusually large variance in copy number in this species,

given this range is almost the same as that reported across 91 different fungal species from

three different fungal phyla (11-251 copies, excluding one outlier of 1442 copies) [5].

However, comparing the *S. cerevisiae* copy number range generated by drawing twelve *S.*

cerevisiae isolates at random from our data 1,000 times to that previously measured for

twelve isolates of one fungal species (*Suillus brevipes*; [5]) shows that the *S. brevipes* range

falls in the middle of the *S. cerevisiae* distribution of copy number ranges (**Fig 7**). These

results suggest that *S. cerevisiae* rDNA copy number is no more variable than that of *S.*

brevipes at least, and illustrate the tremendous variation in rDNA copy number that is likely

to be present in many eukaryotic species.

Figure 7. Distribution of rDNA copy number for fungal and *S. cerevisiae* isolates. The

main histogram represents rDNA copy number (x-axis) for 91 previously published fungal

taxa (blue bars, y-axis on left; [5]) and the 788 *S. cerevisiae* isolates (orange bars; y-axis on

right) from this study. Brown represents overlaps. Inset histogram shows the distribution of

total rDNA copy number ranges from 1,000 randomly drawn sets of twelve *S. cerevisiae*

isolates. The red vertical line represents the total copy number range (84) observed

amongst twelve *Suillus brevipes* isolates [5].

Conclusions

Our results demonstrate that modal coverage can be used to robustly determine rDNA copy

number from NGS data. Using our novel approach, we demonstrate that the mean rDNA

copy number across all wild *S. cerevisiae* populations is 92. This is substantially lower than

the copy numbers documented for lab *S. cerevisiae* strains, but overlaps the 'typical' rDNA

copy numbers reported for fungi [5]. We show that *S. cerevisiae* populations have different

homeostatic rDNA copy numbers, consistent with a previous study [14]. We found some

correlation between rDNA copy number and phylogeny, but not enough to suggest that

homeostatic copy number is simply drifting apart with increasing phylogenetic distance.

We provide circumstantial evidence that environmental factors might help drive the

homeostatic rDNA copy number differences. This is consistent with demonstrations that

nutritional factors can induce physiological rDNA copy number changes [65, 66] and that

such differences have phenotypic consequences [8, 12, 17, 22, 27-35]. However, it has been

shown that rDNA copy number does not correlate with trophic mode in fungi [5].

Therefore, more work is required to determine what really drives copy number dynamics

between populations. One caveat to our conclusions is that while studies from a variety of

organisms have demonstrated that copy number recovers from perturbation [7-10],

presumably as a result of mechanisms maintaining homeostatic copy number [26], some

recent studies in *S. cerevisiae* and *Drosophila* have reported the persistence of stochastic

copy number changes without recovery [65, 67]. It will be important to reconcile these

conflicting results and to determine to what extent the population-level differences we

observe are the result of copy number homeostasis (as we interpret them) versus copy

number inertia.

Our results showing population-level differences in rDNA copy number suggest that such

differences can arise relatively quickly in evolutionary time, although the very high level of

copy number variation between individuals obscures this pattern. Therefore, it is

important to take the large variances and rapid copy number dynamics of the rDNA into

account when interpreting the impact of copy number variation on phenotype.

Bioinformatics pipelines, such as the one we have developed here, in conjunction with the

increasing availability of appropriate NGS datasets provide a way to establish baseline data

on rDNA copy number variation between cells, individuals, populations, and species, as

well as to investigate the phenotypic consequences of this variation. Finally, while we

report population-level differences in rDNA copy number in *S. cerevisiae*, diverse human

populations have been reported to not differ in rDNA copy number [12, 46]. Whether this

reflects a difference in biology (such as differences in the level of genetic divergence

between populations) or an incomplete understanding of human population rDNA copy

number will require further clarification.

Acknowledgements

We thank Gianni Liti (IRCAN, CNRS, INSERM, Université Côte d'Azur, University of Nice) for

kindly providing strains. We acknowledge the Centre for eResearch (CER) and

New Zealand eScience Infrastructure (NeSI) for providing high performance computing

facilities, consulting support, and training services. We thank Auckland Genomics for

advice and whole genome sequencing, and Kevin Chang for help with statistical analyses.

We thank Alastair Harris for helpful suggestions, and the Ganley Lab for helpful comments

and feedback on the manuscript. This work was supported by a grant from the New

Zealand Marsden Fund (14-MAU-053) and a University of Auckland Faculty Research

Development Grant (3712288), both to ARDG.

References

References

1.   Long EO, Dawid IB. Repeated genes in eukaryotes. Annu Rev Biochem. 1980;49:727-

64.

2.   McStay B, Grummt I. The epigenetics of rRNA genes: From molecular to chromosome

biology. Annu Rev Cell Dev Biol. 2008;24:131-57.

3.   Prokopowich CD, Gregory TR, Crease TJ. The correlation between rDNA copy

number and genome size in eukaryotes. Genome. 2003;46:48-50.

4.   Torres-Machorro AL, Hernández R, Cevallos AM, López-Villasenor I. Ribosomal RNA

genes in eukaryotic microorganisms: witnesses of phylogeny? FEMS Microbiol Rev.

2010;34:59–86.

5.   Lofgren LA, Uehling JK, Branco S, Bruns TD, Martin F, Kennedy PG. Genome-based

estimates of fungal rDNA copy number variation across phylogenetic scales and ecological

lifestyles. Mol Ecol. 2019;28(4):721-30. Epub 2018/12/26. doi: 10.1111/mec.14995.

PubMed PMID: 30582650.

6.   Iida T, Kobayashi T. How do cells count multi-copy genes?: "Musical Chair" model for

preserving the number of rDNA copies. Curr Genet. 2019;65(4):883-5. Epub 2019/03/25.

doi: 10.1007/s00294-019-00956-0. PubMed PMID: 30904990.

7.   Kobayashi T, Heck DJ, Nomura M, Horiuchi T. Expansion and contraction of

ribosomal DNA repeats in *Saccharomyces cerevisiae*: requirement of replication fork

blocking (Fob1) protein and the role of RNA polymerase I. Genes and Development.

1998;12:3821-30.

8.   Hawley RS, Marcus CH. Recombinational controls of rDNA redundancy in *Drosophila*.

Annu Rev Genet. 1989;23:87-120.

9.   Russell PJ, Rodland KD. Magnification of rRNA gene number in a *Neurospora crassa*

strain with a partial deletion of the nucleolus organizer. Chromosoma. 1986;93:337-40.

10.   Rodland KD, Russell PJ. Regulation of ribosomal RNA cistron number in a strain of

Neurospora crassa with a duplication of the nucleolus organizer region. Biochimica et

Biophysica Acta. 1982;697:162-9.

11.   Lyckegaard EM, Clark AG. Ribosomal DNA and Stellate gene copy number variation

on the Y chromosome of *Drosophila melanogaster*. PNAS. 1989;86(6):1944-8. Epub

1989/03/01. doi: 10.1073/pnas.86.6.1944. PubMed PMID: 2494656; PubMed Central

PMCID: PMCPMC286821.

12.   Gibbons JG, Branco AT, Yu S, Lemos B. Ribosomal DNA copy number is coupled with

gene expression variation and mitochondrial abundance in humans. Nature

Communications. 2014;5:4850. Epub 2014/09/12. doi: 10.1038/ncomms5850. PubMed

PMID: 25209200.

13.   Cowen LE, Sanglard D, Calabrese D, Sirjusingh C, Anderson JB, Kohn LM. Evolution of

drug resistance in experimental populations of *Candida albicans*. J Bacteriol.

2000;182:1515-22.

14.   West C, James SA, Davey RP, Dicks J, Roberts IN. Ribosomal DNA sequence

heterogeneity reflects intraspecies phylogenies and predicts genome structure in two

contrasting yeast species. Syst Biol. 2014;63(4):543-54. Epub 2014/04/01. doi:

10.1093/sysbio/syu019. PubMed PMID: 24682414; PubMed Central PMCID:

PMCPMC4055870.

15.   Herrera ML, Vallor AC, Gelfond JA, Patterson TF, Wickes BL. Strain-dependent

variation in 18S ribosomal DNA Copy numbers in *Aspergillus fumigatus*. J Clin Microbiol.

2009;47(5):1325-32. Epub 2009/03/06. doi: 10.1128/JCM.02073-08. PubMed PMID:

19261786; PubMed Central PMCID: PMCPMC2681831.

16.   Stults DM, Killen MW, Pierce HH, Pierce AJ. Genomic architecture and inheritance of

human ribosomal RNA gene clusters. Genome Res. 2008;18:13-8.

17.   Ide S, Miyazaki T, Maki H, Kobayashi T. Abundance of ribosomal RNA gene copies

maintains genome integrity. Science. 2010;327:693-6.

18.   French SL, Osheim YN, Cioci F, Nomura M, Beyer AL. In exponentially growing

Saccharomyces cerevisiae cells, rRNA synthesis is determined by the summed RNA

polymerase I loading rate rather than the number of active genes. Mol Cell Biol.

2003;23:1558-68.

19.   Kobayashi T, Ganley ARD. Recombination regulation by transcription-induced

cohesin dissociation in rDNA repeats. Science. 2005;309:1581-4.

20.   Szostak JW, Wu R. Unequal crossing over in the ribosomal DNA of *Saccharomyces*

cerevisiae. Nature. 1980;284(3 Apr):426-30.

21.   Ganley ARD, Kobayashi T. Monitoring the rate and dynamics of concerted evolution

in the ribosomal DNA repeats of *Saccharomyces cerevisiae* using experimental evolution.

Mol Biol Evol. 2011;28:2883-91. Epub 2011/05/07. PubMed PMID: 21546356.

22.   Salim D, Gerton JL. Ribosomal DNA instability and genome adaptability.

Chromosome Research. 2019;27(1-2):73-87. Epub 2019/01/04. doi: 10.1007/s10577-018-

9599-7. PubMed PMID: 30604343.

23.   Ganley ARD, Kobayashi T. Highly efficient concerted evolution in the ribosomal DNA

repeats: total rDNA repeat variation revealed by whole-genome shotgun sequence data.

Genome Res. 2007;17:184-91.

24.   Eickbush TH, Eickbush DG. Finely orchestrated movements: evolution of the

ribosomal RNA genes. Genetics. 2007;175:477-85.

25.   James SA, O'Kelly MJT, Carter DM, Davey RP, van Oudenaarden A, Roberts IN.

Repetitive sequence variation and dynamics in the ribosomal DNA array of *Saccharomyces*

cerevisiae as revealed by whole-genome resequencing. Genome Res. 2009;19:626-35.

26.   Iida T, Kobayashi T. RNA polymerase I activators count and adjust ribosomal RNA

gene copy number. Mol Cell. 2019;73(4):645-54. Epub 2019/01/08. doi:

10.1016/j.molcel.2018.11.029. PubMed PMID: 30612878.

27.   Delany ME, Muscarella DE, Bloom SE. Effects of rRNA gene copy number and

nucleolar variation on early development: inhibition of gastrulation in rDNA-deficient chick

embryos. J Hered. 1994;85(3):211-7. Epub 1994/05/01. doi:

10.1093/oxfordjournals.jhered.a111437. PubMed PMID: 8014461.

28.   Kobayashi T. Regulation of ribosomal RNA gene copy number and its role in

modulating genome integrity and evolutionary adaptibility in yeast. Cell Mol Life Sci.

2011;68:1395-403.

29.   Paredes S, Maggert KA. Ribosomal DNA contributes to global chromatin regulation.

PNAS. 2009;106:17829-34.

30.   Paredes S, Branco AT, Hartl DL, Maggert KA, Lemos B. Ribosomal DNA deletions

modulate genome-wide gene expression: "rDNA-sensitive" genes and natural variation.

PLoS Genet. 2011;7:e1001376.

31.   Michel AH, Kornmann B, Dubrana K, Shore D. Spontaneous rDNA copy number

variation modulates Sir2 levels and epigenetic gene silencing. Genes and Development.

2005;19:1199-210.

32.   Bughio F, Maggert KA. The peculiar genetics of the ribosomal DNA blurs the

boundaries of transgenerational epigenetic inheritance. Chromosome Research.

2019;27(1-2):19-30. doi: 10.1007/s10577-018-9591-2. PubMed PMID: 30511202; PubMed

Central PMCID: PMCPMC6393165.

33.   Cullis CA. Quantitative variation of ribosomal RNA genes in flax genotrophs.

Heredity. 1979;42:237-46.

34.   Xu B, Li H, Perry JM, Singh VP, Unruh J, Yu Z, et al. Ribosomal DNA copy number loss

and sequence variation in cancer. PLoS Genet. 2017;13(6):e1006771. Epub 2017/06/24.

doi: 10.1371/journal.pgen.1006771. PubMed PMID: 28640831; PubMed Central PMCID:

PMCPMC5480814.

35.   Zhou J, Sackton TB, Martinsen L, Lemos B, Eickbush TH, Hartl DL. Y chromosome

mediates ribosomal DNA silencing and modulates the chromatin state in *Drosophila*. PNAS.

2012;109(25):9941-6. Epub 2012/06/06. doi: 10.1073/pnas.1207367109. PubMed PMID:

22665801; PubMed Central PMCID: PMCPMC3382510.

36. Ritossa FM, Spiegelman S. Localization of DNA complementary to ribosomal RNA in

the nucleolus organizer region of *Drosophila melanogaster*. PNAS. 1965;53:737-45. Epub

1965/04/01. doi: 10.1073/pnas.53.4.737. PubMed PMID: 14324529; PubMed Central

PMCID: PMCPMC221060.

37. Wallace H, Birnstiel ML. Ribosomal cistrons and the nucleolar organizer. Biochimica

et Biophysica Acta. 1966;114(2):296-310. Epub 1966/02/21. doi: 10.1016/0005-

2787(66)90311-x. PubMed PMID: 5943882.

38. Schweizer E, MacKechnie C, Halvorson HO. The redundancy of ribosomal and

transfer RNA genes in *Saccharomyces cerevisiae*. J Mol Biol. 1969;40:261-77.

39. Matsuda K, Siegel A. Hybridization of plant ribosomal RNA to DNA: the isolation of a

DNA component rich in ribosomal RNA cistrons. PNAS. 1967;58(2):673-80. Epub

1967/08/01. doi: 10.1073/pnas.58.2.673. PubMed PMID: 5234327; PubMed Central

PMCID: PMCPMC335687.

40. Maleszka R, Clark-Walker GD. Yeasts have a four-fold variation in ribosomal DNA

copy number. Yeast. 1993;9:53-8.

41. Saka K, Takahashi A, Sasaki M, Kobayashi T. More than 10% of yeast genes are

related to genome stability and influence cellular senescence via rDNA maintenance.

Nucleic Acids Res. 2016;44(9):4211-21. Epub 2016/02/26. doi: 10.1093/nar/gkw110.

PubMed PMID: 26912831; PubMed Central PMCID: PMCPMC4872092.

42.   Paredes S, Maggert KA. Expression of *I-Cre*I endonuclease generates deletions within

the rDNA of Drosophila. Genetics. 2009;181:1661-71.

43.   Chestkov IV, Jestkova EM, Ershova ES, Golimbet VE, Lezheiko TV, Kolesina NY, et al.

Abundance of ribosomal RNA gene copies in the genomes of schizophrenia patients.

Schizophrenia Research. 2018;197:305-14. Epub 2018/01/18. doi:

10.1016/j.schres.2018.01.001. PubMed PMID: 29336872.

44.   LeRiche K, Eagle SH, Crease TJ. Copy number of the transposon, *Pokey*, in rDNA is

positively correlated with rDNA copy number in *Daphnia obtuse*. PLoS One.

2014;9(12):e114773. Epub 2014/12/10. doi: 10.1371/journal.pone.0114773. PubMed

PMID: 25490398; PubMed Central PMCID: PMCPMC4260951.

45.   Son J, Hannan KM, Poortinga G, Hein N, Cameron DP, Ganley ARD, et al. rDNA

chromatin activity status as a biomarker of sensitivity to the RNA polymerase I

transcription inhibitor CX-5461. Frontiers in Cell and Developmental Biology. 2020;8:568.

46.   Valori V, Tus K, Laukaitis C, Harris DT, LeBeau L, Maggert KA. Human rDNA copy

number is unstable in metastatic breast cancers. Epigenetics. 2020;15(1-2):85-106. Epub

2019/07/30. doi: 10.1080/15592294.2019.1649930. PubMed PMID: 31352858; PubMed

Central PMCID: PMCPMC6961696.

47.   Alanio A, Sturny-Leclere A, Benabou M, Guigue N, Bretagne S. Variation in copy

number of the 28S rDNA of *Aspergillus fumigatus* measured by droplet digital PCR and

analog quantitative real-time PCR. J Microbiol Methods. 2016;127:160-3. Epub

2016/06/19. doi: 10.1016/j.mimet.2016.06.015. PubMed PMID: 27316653.

48.   Salim D, Bradford WD, Freeland A, Cady G, Wang J, Pruitt SC, et al. DNA replication

stress restricts ribosomal DNA copy number. PLoS Genet. 2017;13(9):e1007006. Epub

2017/09/16. doi: 10.1371/journal.pgen.1007006. PubMed PMID: 28915237; PubMed

Central PMCID: PMCPMC5617229.

49.   Rosato M, Kovarik A, Garilleti R, Rossello JA. Conserved organisation of 45S rDNA

sites and rDNA gene copy number among major clades of early land plants. PLoS One.

2016;11(9):e0162544. Epub 2016/09/14. doi: 10.1371/journal.pone.0162544. PubMed

PMID: 27622766; PubMed Central PMCID: PMCPMC5021289.

50.   Xu J, Xu Y, Yonezawa T, Li L, Hasegawa M, Lu F, et al. Polymorphism and evolution of

ribosomal DNA in tea (*Camellia sinensis*, Theaceae). Mol Phylogen Evol. 2015;89:63-72.

Epub 2015/04/15. doi: 10.1016/j.ympev.2015.03.020. PubMed PMID: 25871774.

51.   Xu J, Zhang Q, Xu X, Wang Z, Qi J. Intragenomic variability and pseudogenes of

ribosomal DNA in stone flounder *Kareius bicoloratus*. Mol Phylogen Evol. 2009;52(1):157-

66. Epub 2009/04/08. doi: 10.1016/j.ympev.2009.03.031. PubMed PMID: 19348952.

52.   Agrawal S, Ganley ARD. Complete sequence construction of the highly repetitive

ribosomal RNA gene repeats in eukaryotes using whole genome sequence data. Methods in

Molecular Biology. 2016;1455:161-81. Epub 2016/09/01. doi: 10.1007/978-1-4939-3792-

9_13. PubMed PMID: 27576718.

53.   Buckler ES, Ippolito A, Holtsford TP. The evolution of ribosomal DNA: Divergent

paralogues and phylogenetic implications. Genetics. 1997;145(March):821-32.

54.   Mayol M, Rosselló JA. Why nuclear ribosomal DNA spacers (ITS) tell different stories

in *Quercus*. Mol Phylogen Evol. 2001;19:167-76.

55.   Wang M, Lemos B. Ribosomal DNA copy number amplification and loss in human

cancers is linked to tumor genetic context, nucleolus activity, and proliferation. PLoS Genet.

2017;13(9):e1006994. Epub 2017/09/08. doi: 10.1371/journal.pgen.1006994. PubMed

PMID: 28880866; PubMed Central PMCID: PMCPMC5605086.

56.   Gong W, Marchetti A. Estimation of 18S gene copy number in marine eukaryotic

plankton using a next-generation sequencing approach. Frontiers in Marine Science.

2019;6:219.

57.   Cubillos FA, Louis EJ, Liti G. Generation of a large set of genetically tractable haploid

and diploid *Saccharomyces* strains. FEMS Yeast Res. 2009;9(8):1217-25. Epub 2009/10/21.

doi: 10.1111/j.1567-1364.2009.00583.x. PubMed PMID: 19840116.

58.   Liti G, Carter DM, Moses AM, Warringer J, Parts L, James SA, et al. Population

genomics of domestic and wild yeasts. Nature. 2009;458(7236):337-41. Epub 2009/02/13.

doi: 10.1038/nature07743. PubMed PMID: 19212322; PubMed Central PMCID:

PMCPMC2659681.

59.   Cox MP, Peterson DA, Biggs PJ. SolexaQA: At-a-glance quality assessment of Illumina

second-generation sequencing data. BMC Bioinformatics. 2010;11:485. Epub 2010/09/30.

doi: 10.1186/1471-2105-11-485. PubMed PMID: 20875133; PubMed Central PMCID:

PMCPMC2956736.

60.   Kumar S, Stecher G, Li M, Knyaz C, Tamura K. MEGA X: Molecular evolutionary

genetics analysis across computing platforms. Mol Biol Evol. 2018;35(6):1547-9. Epub

2018/05/04. doi: 10.1093/molbev/msy096. PubMed PMID: 29722887; PubMed Central

PMCID: PMCPMC5967553.

61.   Keck F, Rimet F, Bouchez A, Franc A. phylosignal: an R package to measure, test, and

explore the phylogenetic signal. Ecology and Evolution. 2016;6(9):2774-80. Epub

2016/04/12. doi: 10.1002/ece3.2051. PubMed PMID: 27066252; PubMed Central PMCID:

PMCPMC4799788.

62.   Pennell MW, Eastman JM, Slater GJ, Brown JW, Uyeda JC, FitzJohn RG, et al. geiger

v2.0: an expanded suite of methods for fitting macroevolutionary models to phylogenetic

trees. Bioinformatics. 2014;30(15):2216-8. Epub 2014/04/15. doi:

10.1093/bioinformatics/btu181. PubMed PMID: 24728855.

63.   Yoon S, Xuan Z, Makarov V, Ye K, Sebat J. Sensitive and accurate detection of copy

number variants using read depth of coverage. Genome Res. 2009;19(9):1586-92. Epub

2009/08/07. doi: 10.1101/gr.092981.109. PubMed PMID: 19657104; PubMed Central

PMCID: PMCPMC2752127.

64.   Peter J, De Chiara M, Friedrich A, Yue JX, Pflieger D, Bergstrom A, et al. Genome

evolution across 1,011 *Saccharomyces cerevisiae* isolates. Nature. 2018;556(7701):339-44.

Epub 2018/04/13. doi: 10.1038/s41586-018-0030-5. PubMed PMID: 29643504; PubMed

Central PMCID: PMCPMC6784862.

65.   Aldrich JC, Maggert KA. Transgenerational inheritance of diet-induced genome

rearrangements in Drosophila. PLoS Genet. 2015;11(4):e1005148. Epub 2015/04/18. doi:

10.1371/journal.pgen.1005148. PubMed PMID: 25885886; PubMed Central PMCID:

PMCPMC4401788.

66.   Jack CV, Cruz C, Hull RM, Keller MA, Ralser M, Houseley J. Regulation of ribosomal

DNA amplification by the TOR pathway. PNAS. 2015;112(31):9674-9. Epub 2015/07/22.

doi: 10.1073/pnas.1505015112. PubMed PMID: 26195783; PubMed Central PMCID:

PMCPMC4534215.

67.   Mansisidor A, Molinar T, Srivastava P, Dartis DD, Pino Delgado A, Blitzblau HG, et al.

Genomic copy-number loss is rescued by self-limiting production of DNA circles. Mol Cell.

2018;72(3):583-93. Epub 2018/10/09. doi: 10.1016/j.molcel.2018.08.036. PubMed PMID:

30293780; PubMed Central PMCID: PMCPMC6214758.