Genomic signatures of innovation and selection in the extremotolerant yeast *Kluyveromyces marxianus*

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Abstract

Extremophiles can be the product of millions of years of evolutionary engineering and refinement. The underlying mechanisms can be quite distinct from the ones operating at earlier stages of trait innovation. In this work, we have developed the compost yeast *Kluyveromyces marxianus,* which diverged from its closest relative >20 million years ago, as a model for interspecies comparative biology and genomics. We applied a battery of growth assays to species of the *Kluyveromyces* genus and found that *K. marxianus* outperformed its relatives in a battery of heat and chemical stress conditions. We then generated and analyzed genomes from across the genus, to find derived genetic features associated with, and potentially causal for, *K. marxianus* traits. We found robust expansions in gene families in the *K. marxianus* genome, most notably among genes annotated as transmembrane transporters and in metabolism. In molecular-evolution tests, we identified adaptive protein variants at hundreds of genes, among which plasma membrane transporters were over-represented. Together, these signals enable a model for the molecular mechanisms and evolutionary pressures underlying *K. marxianus* traits, including gains in transporter function mediating stress resistance, and metabolic variants contributing to its capacity for rapid growth in challenging conditions. Such oligogenic architectures may be the rule rather than the exception in phenotypes that have evolved over long timescales.

Introduction

A central goal of evolutionary biology is to understand trait diversity in the natural world. In many cases, a character of ecological or industrial interest defines a given eukaryotic species, distinguishing it from its relatives. Elucidating the mechanisms of any such phenotype across

species boundaries poses unique technical challenges. As a consequence, against a backdrop

of landmark case studies mapping genotype to phenotype over deep trait divergences (Weiss

and Brem 2019; Smith et al. 2020; Masly and Azom 2022), principles remain at a premium.

What kinds of phenotypes tend to fix in species as they split from one another? How complex is

the genetics, and what are the causal genes and variants? Addressing these questions requires

detailed study of the physiology of ancient traits and their mechanisms.

Single-celled fungi can serve as a powerful model for evolutionary genetics. *Kluyveromyces,* a

genus of budding yeasts, are well-suited to this paradigm given their ~25-million-year

divergence (Shen et al. 2018) and wide-ranging ecology. *Kluyveromyces* have been isolated

from a range of terrestrial and aquatic environments but are best characterized for roles in food

and beverage fermentations (Kurtzman et al. 2011), including elegant evolutionary analyses of

the origin of lactases in dairy-associated lineages (Varela, Puricelli, Ortiz-Merino, et al. 2019;

Friedrich et al. 2023). An extensive observational literature has focused on the growth of one

species in the genus, *K. marxianus,* at temperatures exceeding 45°C, motivated by applications

in the bioproduction industry (Lane et al. 2011; Lertwattanasakul et al. 2013; Mejía-Barajas et

al. 2017; Madeira-Jr and Gombert 2018; Fu et al. 2019; Hoshida et al. 2020; Wu et al. 2020;

Kosaka et al. 2022; Flores-Cosío et al. 2024). The evolutionary driver for the latter trait is

unknown, though *K. marxianus* has been isolated from self-heating agricultural and domestic

compost (Suzuki et al. 2014; Avchar et al. 2022; Verma et al. 2023). Likewise, mechanisms

underlying unique *K. marxianus* traits have remained incompletely understood (Nespolo et al.

2020); potential clues have emerged from comparison of this species to distantly related yeasts,

including reports of resistance to chemical stress (Saini et al. 2017; Linder 2018; Illarionov et al.

2021) and expansions in sugar and potassium importer genes (Varela, Puricelli, Montini, et al.

2019; Papouskova et al. 2024).

We set out to use a comparative approach to trace in detail how and why *K. marxianus* diverged

from its relatives. We first catalogued stress resistance characters that distinguished *K.*

marxianus from the rest of the species of its clade. We then applied genome-content and

molecular-evolution analyses to find unique genomic features in *K. marxianus*. The results

provided insights into the mechanisms and evolutionary forces driving the *K. marxianus* trait

syndrome.

Results

***K. marxianus* exhibits a stress resistance syndrome unique in its genus**

With the ultimate goal of understanding evolutionary innovations in *Kluyveromyces*, we first

sought to survey stress resistance traits across the genus. We profiled the growth in liquid

culture of four species associated with terrestrial habitats (*K. marxianus, K. lactis, K.*

wickerhamii, and *K. dobzhanskii*) and three aquatic species (*K. aestuarii*, *K. siamensis*, and *K.*

nonfermentans) in a panel of media conditions (Figure 1 and Table S1). At 42°C, *K. marxianus*

exhibited a qualitative growth advantage over its relatives, to an extent far beyond the pattern in

a permissive control condition (rich medium at 28°C; Figure 1B); genetically distinct *K.*

marxianus isolates phenocopied each other with respect to high-temperature growth (Figure

S2), as expected (Lane et al. 2011). In ethanol, caffeine, the DNA intercalator propidium iodide,

and the DNA damage agent methyl methanesulfonate (MMS), *K. marxianus* likewise

outperformed the other species of the genus (Figure 1B). Our profiling also identified other

conditions in which *K. marxianus* did not stand out with respect to growth (Figure S1). We

conclude that *K. marxianus* is unique in the genus for its resistance to a broad subset of abiotic

challenges, including high temperature.

K. marxianus resistance phenotypes are unique for their dependence on growth phase but not oxygen

For many microbes, when cells face nutrient exhaustion, they shut down exponential (log-

phase) growth and enter the quiescent, extra-durable state called stationary phase. The

difference between _K. marxianus_ and its relatives in stress resistance that we had noted across

a growth timecourse (Figure 1) might also manifest in the stationary state or, alternatively, its

mechanism could be tied to active cell division. To distinguish between these possibilities, we

focused on thermotolerance as a test case. We used a viability assay in which we inoculated a

given liquid culture into rich medium; incubated for growth at the permissive temperature until

nutrient exhaustion; subjected this stationary-phase culture to a short heat shock; and used it to

seed colonies for growth at the permissive temperature on solid medium (Figure 2A). Results

revealed only modest differences between species in heat shock resistance in this stationary-

phase heat treatment paradigm, with no particular advantage for _K. marxianus_ (Figure 2B). By

contrast, when we treated log-phase cultures with heat shock, _K. marxianus_ exhibited much

higher viability than did the other species of the genus (Figure 2C), paralleling its advantage

during chronic high-temperature exposure (Figure 1B). No such differences were apparent in a

permissive 28°C control experiment (Figure S3). We conclude that for high temperature, the _K._

marxianus advantage only manifests when cells are actively dividing, and that cell death, rather

than slow or arrested growth, mediates the defects of the rest of the genus.

To extend our analysis of cell death as the underlying mechanism of species differences in

stress, we next assayed viability in log-phase cultures across a panel of representative

stresses, focusing on the stress-sensitive species _K. lactis_ as a point of comparison for _K._

marxianus (Figure 1). The latter exhibited a survival advantage in each stress condition (Figure

S4). The difference manifested regardless of oxygen availability, though both species performed

better in normoxia (Figure S4), as expected (Lertwattanasakul et al. 2013; Mejía-Barajas et al.

2017; Madeira-Jr and Gombert 2018; Fu et al. 2019; Hoshida et al. 2020; Kosaka et al. 2022).

Together, these trends from assays of viability establish the resistance character of *K.*

marxianus as a lineage-unique gain of mechanisms that protect cells from lethal failures of the

proliferation program under stress.

Transcriptional remodeling by *K. lactis*, but not *K. marxianus*, in stress

To learn more about the state of *Kluyveromyces* cells under stress, we turned to an expression

profiling approach, for which we again chose *K. lactis* as a model stress-sensitive species for

comparison against *K. marxianus*. We surveyed the transcriptomes of the two species in

representative treatments where we had observed divergent growth phenotypes—high

temperature, caffeine, MMS, and ethanol—and in the permissive control of rich medium at 28°C

(Table S2). In principal component analyses, the strongest contributor to variation across the

data set was a program of housekeeping genes expressed more highly in *K. lactis* than *K.*

marxianus in all conditions, including controls (Figure 3A, Figure S5A, and Table S2). Also

salient on a global scale was a suite of genes that *K. lactis* turned up and down depending on

treatment; the *K. marxianus* transcriptome was much less volatile (Figure 3A, Figure S5A, and

Table S2).

For a more fine-grained analysis, we tested Gene Ontology groups for coherent, stress-evoked

expression change (Table S3). The results revealed housekeeping gene groups induced in

stress by *K. lactis*, including ribosomal RNA processing and protein folding (Figure 3B, Figure

S5B, and Table S3). Additionally, in a few cases, regulons in *K. lactis* and *K. marxianus*

paralleled each other, notably repression of glycolysis and induction of the electron transport

chain in some stresses (Figure 3C and Figure S5C). Taken together, however, our

transcriptional analyses established programs of elevated expression by *K. lactis*—in stress and

in untreated controls—as its major regulatory distinction from *K. marxianus*. Such a pattern

would be expected if *K. lactis* underwent more functional failures during growth in our panel of

conditions, setting off compensatory expression responses downstream.

Unique transporter and metabolic gene duplications in the *K. marxianus* genome

In continued pursuit of mechanisms underlying traits that distinguished *K. marxianus* from its

relatives, we reasoned that unique genomic features in this species could serve as a source for

potential hypotheses. As a first strategy to find genomic novelties in *K. marxianus*, we focused

on species-specific gene duplications in paralogous gene families. From annotated reference

genomes for six *Kluyveromyces* species (*K. marxianus, K. lactis, K. wickerhamii*, *K.*

dobzhanskii, *K. aestuarii*, and *K. nonfermentans*; Figure 1A), plus outgroups from nearby clades

in the Saccharomycotina subphylum, we identified gene families and the members detectable in

each species in turn with the CAFE 5 package (Mendes et al. 2021). This approach identified 16

gene families with inferred gain or loss events along the *K. marxianus* lineage at rates that were

unusual relative to the average across the species set (Figure 4A and Table S4), including

families subject to tandem duplications and those with members across the *K. marxianus*

chromosomes (Figure 4B and Table S4). Of these, 13 were families with extra copies in *K.*

marxianus that could be classified based on domain annotations as functioning in metabolism or

plasma membrane transport, the latter including both nutrient importers and xenobiotic efflux

pumps (Figure 4A). These data make clear that gene family expansions are prevalent in the *K.*

marxianus genome, providing a line of genomic evidence that highlights small-molecule

transport and metabolism as candidate mechanisms of *K. marxianus* phenotypes.

Widespread evidence for positive selection on amino acid alleles in *K. marxianus*

As a further means to identify genomic correlates of *K. marxianus* phenotypes, we pursued

molecular-evolution analyses of amino acid variation on a per-gene basis. We reasoned that

some phenotypes unique to *K. marxianus* could have evolved as a product of adaptation to

abiotic stress in challenging ecological niches (including in compost), and that at relevant loci,

we could detect evidence for positive selection on protein-coding sequences. For a phylogenetic

test toward this end, at each gene in turn, we aligned orthologs from reference genomes for our

six focal *Kluyveromyces* species, to which we fit a model in the PAML package (Yang 2007)

specifying positive selection along the *K. marxianus* lineage for some codons, with all other

codons and branches evolving under selective constraint. This model was justified, over and

above a null model with no selection, for a remarkable 525 genes (Figure 5A, 5C, and 5D and

Table S5). By contrast, controls focused on the *K. lactis* branch identified just five genes with

signal for positive selection (Figure 5A and Table S5), supporting the inference of a history of

particularly avid adaptation at the amino acid level in *K. marxianus*.

For most gene hits from our PAML branch-site tests focused on *K. marxianus*, a small number

of codons per gene were classified as targets of positive selection (Figure 5B). We expected

that, if the derived alleles at each such site had undergone a selective sweep through *K.*

marxianus populations, we could detect it across strains of this species. To test this, we carried

out short-read resequencing of the genomes of five wild *K. marxianus* isolates (Table S1) and

used the resulting reads alongside our reference genome of the Km31 *K. marxianus* strain as

input to a pipeline for variant calling, pseudogenome construction, and sequence alignment. We

then inspected these alignments at each site where the *K. marxianus* allele had arisen under

selection according to our analysis of reference genomes with PAML. At 99% of the latter, we

detected complete conservation across *K. marxianus* strains (Figure 5C and Table S5). These

phylogenetic and resequencing data are consistent with the scenario of positive selection in *K.*

marxianus in hundreds of proteins.

Enrichment for amino acid variation between *Kluyveromyces* species in transporter

genes

To complement our phylogenetic analyses, we next applied population-based molecular-

evolution tests that used our *K. marxianus* resequencing data and a population-genomic

resource for *K. lactis* (Friedrich et al. 2023). We implemented McDonald-Kreitman molecular

evolution paradigms (McDonald and Kreitman 1991) to assess evidence for differences in

selective pressure between the species on a given protein sequence, as follows. For each

coding region we tabulated sites of non-synonymous and synonymous divergence between *K.*

marxianus and *K. lactis*, and likewise for polymorphic variants within the species (Table S6).

Using these counts in a McDonald-Kreitman test on a per-gene basis, we identified 11 genes

with a significant excess of interspecies amino acid divergence across the coding region

(relative to polymorphism and for synonymous variation; Figure 5D and Table S6). Thus,

relative to the branch-site phylogenetic method (Figure 5A), the McDonald-Kreitman scheme

was under-powered in this system, as expected if most loci only had a few amino acid targets of

adaptation between *K. marxianus* and *K. lactis* (Figure 5B). But among the gene hits that did

emerge from the McDonald-Kreitman test, each represented an inference of accelerated protein

divergence between the species that complemented our phylogenetic results; four such loci had

also been top-scoring in our phylogenetic framework, two annotated in metabolism and two in

transmembrane transport (Figure 5C and 5D).

Separately, we developed a genomic version of the McDonald-Kreitman paradigm, in which we

tabulated the neutrality index (NI) for a given gene as non-synonymous divergence normalized

for polymorphism and for synonymous variation, and we evaluated significance of this metric

across cohorts of genes of similar function via resampling (Figure 5E and Table S7). For

metabolic gene groups with extreme NIs in this test, focused analyses indicated that the signal

was driven by low polymorphism, reflecting a history of purifying selection (Table S7). By

contrast, for a group of transmembrane transporters, the individual components of the NI metric

followed the pattern indicative of *bona fide* positive selection: non-synonymous variation

between species was high in the genes of the group and polymorphism levels were on par with

the genomic average (Figure 5E and Table S7). This selection signal in transporter genes as a

group dovetailed with our observation of individual transporters emerging as top-scoring in

single-gene McDonald-Kreitman tests as phylogenetic analyses (Figure 5D); it also echoed our

observation of transporter gene family expansions in *K. marxianus* (Figure 4). As such, our

complete set of genomic analyses attest to a complex history of selection on loci throughout the

K. marxianus genome, and they uncover innovations in transporter function as a particularly

compelling candidate mechanism of unique traits in this species.

Discussion

In this work, we have established *Kluyveromyces* as a genomic and genetic model for the study

of species-unique traits. Our results have revealed attributes that distinguish *K. marxianus* from

the rest of its genus: stress resistance in growth assays, expression stability in the face of

environmental change, and sequence evidence for gene-content and amino-acid variation.

These observations shed new light on the biology and evolution of *K. marxianus*,

complementing classic trait compendia (Kurtzman et al. 2011) and the modern literature (Lane

and Morrissey 2010; Lane et al. 2011; Madeira-Jr and Gombert 2018). In terms of genetics, the

alleles we have found to be unique in *K. marxianus* represent candidates for the components

that nature would have used to build the trait syndrome in this species and that would be

sufficient to reconstitute it in an exogenous background—as opposed to the handful of previous

studies that have reported genes necessary for thermotolerance in *K. marxianus* (Falcão

Moreira et al. 1998; Lertwattanasakul et al. 2013; Wu et al. 2020; Montini et al. 2022).

Most pervasive in our genomic sequence scans was the trend of expanded transmembrane

transporter gene families, and new amino acid alleles in single-copy transporter genes, in the *K.*

marxianus genome relative to the rest of its clade. These include gains by *K. marxianus* in

sugar and potassium transporter repertoires, which were previously analyzed in comparisons

with distant fungal relatives (Varela, Puricelli, Montini, et al. 2019; Papouskova et al. 2024) and

which we can now date to the divergence of *K. marxianus* from its sister species. Our results

also uncovered alleles unique to *K. marxianus* in many other ion and nutrient importers that

function at the plasma membrane, as well as xenobiotic exporters and the transporter regulatory

machinery called the eisosome. Such changes could contribute to the stress resistance

phenotypes we have seen in *K. marxianus* if they improve the selectivity of nutrient importers or

the avidity of toxin exporters. Alternatively, transporter changes in *K. marxianus* could represent

compensation for a separate, as yet unknown heightened barrier function at the plasma

membrane and/or cell wall: any such extra protection could boost stress resistance at the

expense of passive nutrient acquisition (Ferenci 2016) and drive compensatory changes to

improve the machinery for active transport.

That said, we observed signals of evolutionary innovation by *K. marxianus* in genes of many

functions beyond transmembrane transport. We thus picture a multifaceted mechanism for

resistance traits in this system. Under one extreme model, the complement of new gene copies

and new amino acid alleles that we have found in *K. marxianus* could work together to mediate

a largely unified, global stress-resistance program. Perhaps more plausibly, subsets of these

variants could have more specific phenotypic roles, if *K. marxianus* evolved different

mechanisms for resistance to different insults. Indeed, hit loci from our analyses of the *K.*

marxianus genome span a much wider range of annotations than the variant cell growth genes

mapped in previous studies of *S. cerevisiae* thermotolerance (Weiss et al. 2018; AlZaben et al.

2021; Abrams et al. 2022)—likely reflecting the evolution of resistance to a bigger swath of

stresses by *K. marxianus.*

If the trait syndrome in *K. marxianus* were mediated by hundreds of unlinked loci, it would

conform to the view of genetic complexity as the rule rather than the exception for trait variation,

across the tree of life (Rockman 2012; Boyle et al. 2017). Even the novel gene copies and

amino acid alleles we have tabulated in the *K. marxianus* genome could represent an

underestimate of the full genetic architecture of traits in this species, given the potential of false

negatives in our approach. In other systems, signal for selection on such a scale has been

reported in the study of major life transitions (*e.g.* the evolution of marine mammals from

terrestrial ancestors (Chikina et al. 2016) and snakes from tetrapods (Castoe et al. 2013)), and

in analyses of convergent traits in species complexes (Smith et al. 2020), from tomato (Pease et

al. 2016) to long-lived fishes (Kolora et al. 2021) and eusocial bees (Woodard et al. 2011). The

ecological pressures on the *K. marxianus* ancestor remain poorly established. Though well-

recognized as an industrial dairy yeast (Prado et al. 2015; Gethins et al. 2016; Ortiz-Merino et

al. 2018; Varela, Puricelli, Ortiz-Merino, et al. 2019; Yazdi et al. 2022), *K. marxianus* was likely

domesticated for this purpose fairly recently, as inferred for *K. lactis* (Friedrich et al. 2023). And

though *K. marxianus* can also be isolated from rotting fruit (Williams et al. 2019; Al-Qaysi et al.

2021), it may not have specialized narrowly to such substrates: a neighboring clade that

includes *Saccharomyces cerevisiae* produces more ethanol than do any of the *Kluyveromyces*

species (Nespolo et al. 2020), suggesting a weaker drive in the latter toward this putative

mechanism to kill off microbial competitors in high-sugar environments. Among microbial-

ecology findings, the strongest clue to the origin of unique phenotypes in *K. marxianus* lies in its

prevalence in decomposing plant material other than fruit, including leaf litter and pine needles

(Spurley et al. 2022), cow dung (Avchar et al. 2022), industrially processed cocoa beans

(Baptista et al. 2021), agave (Chacón-Vargas et al. 2020), and sugar cane (Anderson et al.

1986; Cernak et al. 2018) as well as industrial and domestic compost (Suzuki et al. 2014;

Avchar et al. 2022; Verma et al. 2023). Plant defense compounds in decomposing material are

a compelling candidate driver for the rise of resistance traits in *K. marxianus* (Baptista and

Domingues 2022) since they can persist for days in a leaf or stem after abscission (Burton

1982; Goodspeed et al. 2013; Shimada et al. 2015). Furthermore, some evidence for low pH

deep in compost piles (Rainey and Covey 2016) suggests that this condition could also have

contributed to *K. marxianus* evolution. In this scenario, the phenotypic syndrome of *K.*

marxianus would involve not only resistance to the stresses of the decaying plant environment,

but also the rapid utilization of simple sugars liberated early in plant breakdown, mediated by

the new copies and new alleles of metabolic genes we have noted in the *K. marxianus* genome.

As *K. marxianus* lacks the capacity to degrade lignocellulosic plant cell walls, these would be

left to other fungal saprotrophs acting later in decomposition.

We thus envision an ecological model in which *K. marxianus* originally evolved as a plant decay

specialist, with extant strains coming to occupy a broad range of niches, including soil and

human and animal material (Rij 2013; Ortiz-Merino et al. 2018; Zhang et al. 2021). Given that

we and others have seen conservation of resistance traits across *K. marxianus* strains (Lane et

al. 2011; Madeira-Jr and Gombert 2018; Lappe-Oliveras et al. 2023), we infer that unique

capacities arose early in *K. marxianus,* sometime after *K. marxianus* originated as a species,

~25 million years ago (Shen et al. 2018), and persisted during its radiation into modern

populations.

Assuming that hundreds of species-unique variants across the genome prove to explain the *K.*

marxianus trait syndrome, they could have accumulated over a very long time. Current theory

(Jain and Stephan 2017; Hayward and Sella 2022) would posit that, along the *K. marxianus*

lineage, a first advantage arose in ecologically relevant stress conditions as a product of a few

major-effect loci, followed later by the accumulation of many more variants with smaller adaptive

effects. Dynamics of this kind, assuming they emerge as generally relevant, could unify the

current understanding of genetically simple, recent, intraspecies (local) adaptations (Bell 2009)

with that of more ancient traits. Empirical validation of this model, in lower and higher

organismal systems, represents an ongoing goal of interspecific genetics.

Methods

1. Strains

Strains used in phenotyping experiments and population-genomic analyses are listed in Table

S1.

2. Liquid growth phenotyping of wild-type *Kluyveromyces*

Liquid culture experiments in stress and standard-condition controls in Figure 1 used *K.*

marxianus Km31 (Cernak et al. 2018); *K. lactis* yLB72 (Sorrells et al. 2015); and the strains of

K. dobzhanskii, K. wickerhamii, K. nonfermentans, K. aestuarii, and *K. siamensis* listed in Table

S1. For a given species and condition, we streaked a frozen stock onto yeast peptone dextrose

(YPD) agar plates, incubated for two days at 28°C, picked a single colony, inoculated into 5 mL

liquid YPD, and incubated with shaking at 28°C for 16-18 hours. We call this the first pre-

culture. We diluted, into fresh liquid YPD, an inoculum of the first pre-culture at the volume

required to attain an optical density (OD_{600}) of 0.05 for a total culture volume of 5 mL, and we

incubated this second pre-culture with shaking at 28°C for ~7 hours until it reached OD_{600} = 0.2.

For unstressed controls in Figure 1, we diluted an inoculum of the second pre-culture into fresh

liquid YPD at the volume required to attain OD_{600} = 0.2 for a total culture volume of 5 mL. 150 µL

of this experimental culture was transferred to each of 12 wells of a 96-well plate, sealed with a

Breathe-Easy membrane (Diversified Biotech), for incubation at 28°C and growth

measurements across a 24-hour timecourse in a Tecan M200 PRO plate reader (parameter

time point measurement, OD_{595}). For heat stress treatment in Figure 1, setup and growth

assays were as above except that incubation of the experimental culture was at 42°C. For

chemical stress treatments in Figure 1, setup and growth assays were as above except that for

the experimental culture, before transfer to the 96-well plate, ethanol (Koptec) was added to

attain a concentration of 6.5% (v/v); caffeine (Sigma-Aldrich) was added to attain 25 mM;

methane methylsulfonate (MMS; Thermo Scientific) was added to attain 0.03% (v/v); or

propidium iodide was added to attain 1mM, and growth timecourses were 48 to 72 hours. For

chemical stress treatments and growth measurements in Figures S1 and S2, for a given

species and condition, streaking and the first pre-culture were as above. We diluted an

inoculum of the first pre-culture into fresh liquid YPD at the volume required to attain OD_{600} = 0.2

for a total culture volume of 5 mL, with 1 M NaCl (Fisher), 3% glycerol (v/v; Fisher), 80 mM

$CaCl_2$ (Sigma), 250 mM rapamycin (Research Products International), 6 mM H_2O_2 (Fisher), or

16% (v/v) Triton X-100 (Fisher), followed by incubation and growth measurements in the 96-well

format as above.

3. Assays of cell survival from *K. marxianus* and *K. lactis* stationary and log-phase liquid

culture

For survival assays of cells in exponential phase in Figures 2 and S3, for *K. marxianus* Km31

and separately for *K. lactis* yLB72 we streaked a frozen stock onto yeast peptone dextrose

(YPD) agar plates, incubated for two days at 28°C, picked a single colony, inoculated into 5 mL

liquid YPD, and incubated with shaking at 28°C for 16-18 hours. We diluted, into fresh liquid

YPD, an inoculum of this first pre-culture at the volume required to attain OD_{600} = 0.05 for a total

culture volume of 5 mL, and we incubated this second pre-culture with shaking at 28°C for ~7

hours until it reached OD_{600} = 0.2. We took this second pre-culture directly into experimental

treatment as follows. For unstressed controls, serial dilutions of the culture were spotted onto

agar plates and incubated for 2 days at 28°C. For heat shock, 1 mL of the culture was incubated

in a heat block at 47°C for 30 minutes and then transferred to ice for 2 minutes, followed by

spotting and incubation as above.

For survival assays of cells in stationary phase in Figures 2 and S3, for *K. marxianus* Km31 and

separately for *K. lactis* yLB72 we streaked a frozen stock onto yeast peptone dextrose (YPD)

agar plates, incubated for two days at 28°C, picked a single colony, inoculated into 5 mL liquid

YPD, and incubated with shaking at 28°C for 96 hours. Heat shock and spotting (and

unstressed controls) were as above.

4. Assays of cell survival from *K. marxianus* and *K. lactis* anaerobic and aerobic liquid

cultures

As *Kluyveromyces* cannot tolerate strictly anaerobic conditions (Kiers et al. 1998; Ishtar Snoek

and Yde Steensma 2006; Merico et al. 2007; Dekker et al. 2021), for Figure S4 we used a

limited-oxygen experimental paradigm (Hoshida et al. 2020) as follows. For *K. marxianus* Km31

and separately for *K. lactis* yLB72, for each treatment we isolated a single colony on solid plates

and carried out first and second pre-cultures in liquid YPD at 28°C under aerobic conditions as

in Section 2. For unstressed controls in anaerobic conditions, we diluted an inoculum of the

second pre-culture into fresh liquid YPD at the volume required to attain OD_{600} = 0.05 for a total

culture volume of 5 mL. 150 µL of this experimental culture was transferred to each of 12 wells

of a 96-well plate, which was placed (with lid) into an AnaeroPouch (ThermoScientific) with a

AnaeroPouch - Anaero gas pack. The sealed pouch and plate were incubated for 72 hours at

28°C. Then we collected 30 µL of culture from each well and combined them all into one master

culture, 100 µL of which we serially diluted into spots on agar plates, followed by incubation for

2 days at 28°C. Aerobic controls were treated analogously except that the 96-well plate was

inoculated without a pouch. Viability assays of chemically stressed cells in anaerobic conditions

and aerobic controls were as above except that for the experimental culture, before transfer to

the 96-well plate, ethanol, caffeine, or MMS were added as in Section 2. Viability assays of

heat-treated cells in anaerobic conditions and aerobic controls were as above except that the

sealed pouch and plate were incubated for 24 hours at 39°C.

5. Transcriptional profiling

5.1. Culture protocol and experimental design. For a given replicate of RNA-seq of *K. marxianus*

Km31 and separately for *K. lactis* yLB72 in unstressed control conditions, we isolated a single

colony on solid plates and carried out first and second pre-cultures in liquid YPD at 28°C as in

Section 2. For unstressed controls, we diluted an inoculum of the second pre-culture into fresh

liquid YPD at the volume required to attain OD_{600} = 0.05 for a total culture volume of 50 mL. This

experimental culture was incubated at 28°C with shaking until it reached OD_{600} = 0.4-0.85. From

this culture we took a volume corresponding to 2 OD units, pelleted its cells by centrifugation for

5 minutes at 1,500 x g, and resuspended in 1 mL of RNAlater (ThermoFisher) followed by

storage at 4°C for up to 48 hours. For RNA-seq of heat-stressed cells, culture, pelleting, and

storage were as above except that the experimental culture was incubated at 37°C or 39°C; for

RNA-seq of cells under chemical stress, culture, pelleting, and storage were as above except

that the experimental culture was done in YPD with 22 mM caffeine, 0.01% MMS, or 3%

ethanol. For expression-based reannotation of *K. marxianus* Km31 and *K. lactis* yLB72 (see

below), we used the above pipeline for a preliminary expression profiling experiment which we

call the training batch: we cultured one biological replicate of Km31 and yLB72 at 37°C and one

of Km31 and yLB72 in unstressed control conditions. For comprehensive transcriptional profiling

analyses in Figure 3 and Figure S5, we used the above pipeline to generate cell pellets in two

production batches. In one batch we grew three biological replicates of Km31 and three of

yLB72 in 39°C heat stress, and three unstressed cultures of each species. In another batch we

grew three biological replicates of each species in caffeine, three in MMS, and three in ethanol,

and three unstressed controls.

5.2. RNA isolation and sequencing. For RNA isolation of a given sample pellet stored in

RNAlater (see Section 5.1), we centrifuged for 5 minutes at 1,500 x g, and after decanting of the

supernatant, we resuspended the pellet in 200 μL of nuclease-free water followed by addition of

750 μL of BashingBead buffer from the Direct-Zol RNA MicroPrep RNA kit (Zymo Research).

We then transferred our 950 μL of suspension into BashingBead tubes, vortexed for 20

minutes, and centrifuged for 5 minutes at 12,000 x g. We took ~600 μL of supernatant from the

top layer being careful not to touch the pellet, transferred to a new tube, and mixed with an

equal volume of 100% ethanol. This mixture was used as input to the rest of the kit, following

the manufacturer's instructions. RNA library preparation and sequencing of ~20M 150 bp

paired-end reads per sample was carried out by Novogene, Inc.

5.3. Genome re-annotation and ortholog calls for K. marxianus *and* K. lactis

To generate a reference genome for *K. marxianus* strain Km31, we proceeded as follows. We

first used bwa (bio-bwa.sourceforge.net) to map the RNA-seq reads from the *K. marxianus*

cultures of the training batch of our transcriptional profiling (see Section 5.1) to a published

genome of the DMB1 strain of *K. marxianus* (Suzuki et al. 2014). We then used these

mappings, requiring mapping qualities of >20 and mapped lengths >20, to call single-nucleotide

polymorphisms with BCFtools (Li 2011; Danecek et al. 2021) mpileup with the -I flag to ignore

indels, and filtering for exclusion of variants with: a depth of less than 10 reads, or more than

10% of reads with a mapping quality of 0 supporting the variant. At the called polymorphisms,

we computationally introduced Km31 alleles into the DMB1 genome sequence with the

consensus tool in BCFTools to generate a first-pass pseudogenome for Km31. Next, we used

this pseudogenome and the training batch RNA-seq reads as input into transcript assembly with

Trinity (Grabherr et al. 2011) and annotation with PASA (Haas et al. 2003), and we took the

output as our final annotated genome for *K. marxianus* Km31 (Supplementary Data 1). We

generated a reference genome for *K. lactis* strain yLB72 as above, except that we used RNA-

seq from the *K. lactis* training batch and the published genome of *K. lactis* strain NRRL Y-1140

(Dujon et al. 2004) (Supplementary Data 1). To call orthologous gene pairs between *K.*

marxianus Km31 and *K. lactis* yLB72, we used our final annotations from the two species as

input into OrthoFinder (Emms and Kelly 2019). We retained all one-to-one orthology calls, *i.e.*

cases in which a single annotated gene from *K. marxianus* was called orthologous to a single

gene from *K. lactis*. We also sought to valorize cases in which OrthoFinder, with the goal of

finding orthologs in species B of a gene $G_{1,A}$ in species A, did not distinguish between n

candidate orthologous genes $G_{1...n,B}$ from species B. We evaluated each candidate $G_{i,B}$ for

synteny: if the next gene up- or downstream of $G_{i,B}$ in the species B genome was called a one-

to-one ortholog of the up- or downstream genes of $G_{1,A}$ in the species A genome, we retained

this case as a bona fide ortholog pair. Our final set comprised 3578 ortholog pairs between *K.*

marxianus Km31 and *K. lactis* yLB72. *Saccharomyces* orthologs were taken from (Sorrells et al.

2018).

5.4. Production-batch transcriptional profiling data analysis

For comprehensive transcriptional profiling analyses in Figure 3 and Figure S5, we used our

RNA-seq data from cultures of *K. marxianus* Km31 and *K. lactis* yLB72 from production batches

(see Section 5.1) as follows. For the RNA-seq reads from a given replicate of a given species

(*K. marxianus* or *K. lactis*) and condition (39°C, ethanol, caffeine, MMS, or unstressed control),

we mapped to the respective pseudogenome (see Section 5.3) using the TopHat (Kim et al.

2013) aligner with default parameters, and we used these mappings as input into HTseq

(Anders et al. 2015) to generate raw read counts per gene. For a given species and stress, we

used the counts from all replicates from the stress, and the unstressed control from the

respective batch, as input into edgeR (Robinson et al. 2010) to generate normalized counts per

gene and replicate and a normalized $\log_2$ fold-change between stress and control conditions as

an average across replicates. Principal component analysis (PCA) of the RNA-seq data was

done using the PCA module of scikit-learn (Pedregosa et al. 2011). The edgeR normalized

counts for each gene in each treatment were scaled to unit variance with scikit-learn before

running the PCA. Principal component weights are listed in Supplementary Table 2.

To identify gene groups in which a given stress treatment evoked significant gene induction or

repression relative to the unstressed control, in *K. marxianus* Km31 and separately in *K. lactis*

yLB72 (Table S3), we carried out resampling-based enrichment tests as follows. We used

assignments of each gene to Gene Ontology terms for *K. lactis* from FungiDB

(https://fungidb.org/fungidb/app/record/dataset/NCBITAXON_284590), and we propagated

these assignments to orthologous genes in *K. marxianus* from Section 5.3. For a given species

and stress, for each Gene Ontology term in turn, for the n genes from the term with expression

observations, if $n > 5$ we collated the genes' $\log_2$ fold-changes in expression between stress

and the untreated control, each of which was a mean across replicates (from edgeR; see

above). We took the mean of these fold-changes across the genes of the term, m_{true}. We then

picked n random genes from the genome and tabulated their mean expression fold-change for

the respective species and stress, $m_{resample}$. Repeating the latter resampling 1000 times, we

calculated an empirical p-value reporting the enrichment of the magnitude of expression fold-

change as the percentage of resamples for which ($|m_{resample}| \geq (|m_{true}|)$. We used the p-values for

all terms tested for the respective species and stress as input into multiple testing correction

with the Benjamini-Hochberg method.

7. Molecular-evolution analyses

7.1. Gene duplication analysis

As input to the gene family copy number analysis for Figure 4 and Table S4, genomes and

annotations were as follows. For *K. marxianus* and *K. lactis* we used the reference genomes

and annotations of Km31 and yLB72, respectively, from Section 5.3. For *K. wickerhamii* we

used the reference genome and annotation of UCD54-210 from (Sorrells et al. 2018). For the *K.*

nonfermentans strain NRRL Y-27343 (GenBank acc. no. GCA_030569915.1), the *K. aestuarii*

strain NRRL YB-4510 (GenBank acc. no. GCA_003707555.2), the *K. siamensis* strain CBS

10860 (GenBank acc. no. GCA_030579315.1), and the *K. dobzhanskii* strain NRRL Y-1974

(GenBank acc. no. GCA_003705805.3) we used YGAP (Proux-Wéra et al. 2012) to generate *de*

novo annotations from previously generated reference genomes (Supplementary Data 1). As

outgroup species we used the genomes and annotations for *Ashbya aceri* (GenBank acc. no.

GCA_000412225.2), *Candida albicans* (GenBank acc. no. GCA_000182965.3), *Eremothecium*

cymbalariae (GenBank acc. no. GCA_000235365.1), *Eremothecium gossypii* (GenBank acc.

no. GCA_000091025.4), *Eremothecium sinecaudum* (GenBank acc. no. GCA_001548555.1),

Lachancea dasiensis (GenBank acc. no. GCA_900074725.1), *Lachancea fantastica* (GenBank

acc. no. GCA_900074735.1), *Lachancea fermentatii* (GenBank acc. no. GCA_900074765.1),

Lachancea lanzarotensis (GenBank acc. no. GCA_000938715.1), *Lachancea meyersii*

(GenBank acc. no. GCA_900074715.1), *Lachancea mirantina* (GenBank acc. no.

GCA_900074745.1), *Lachancea nothofagi* (GenBank acc. no. GCA_900074755.1), *Lachancea*

quebecensis (GenBank acc. no. GCA_002900925.1), *Lachancea thermotolerans* (GenBank

acc. no. GCA_000142805.1), and *Saccharomyces cerevisiae* (GenBank acc. no.

GCA_000146045.2). We used proteomes for each species as input into OrthoFinder (Emms

and Kelly 2019) to find groups of orthologs. We then used the time-weighted tree and gene

family counts from OthoFinder as input into CAFE 5 (Mendes et al. 2021) to analyze gene

family contractions and expansions. Gene families that did not exist at the root were included.

We adjusted *p*-values with the Benjamini-Hochberg test to correct for multiple testing. We used

CafePlotter (https://github.com/moshi4/CafePlotter) to visualize significant gene family

expansions and contractions on the phylogenetic tree. We then assigned functions for each

gene family by running all proteins from all species for each gene family through InterProScan

(Jones et al. 2014). We took the most common InterPro term in the output of the scan as the

inferred function of the gene family in Figure 4.

7.2. Phylogenetic analyses

As input to phylogenetic analyses of one strain for each *Kluyveromyces* species for Figure 5

and Table S5, we used genomes and annotations for the *Kluyveromyces* genus excluding *K.*

siamensis as in Section 7.1. We called orthologs across these six species with OrthoFinder

(Emms and Kelly 2019), and for a given orthogroup we generated protein alignments using

MUSCLE (Edgar 2004). We then used this alignment and its nucleotide sequences as input into

PAL2NAL (Suyama et al. 2006) using the -nogap flag to remove gaps to generate a codon

alignment. For each orthogroup in turn, we ran the PAML (Yang 2007) codeml package run

using the branch-site model (Zhang et al. 2005) to evaluate the fit to the sequence data of a

model in which regions of the coding sequence are under positive selection for amino acid

variation along a focal (foreground) branch and not on any other branch from the tree (the

background branches). We used as input to PAML the codon alignment for the orthogroup and

the topology of the phylogenetic tree relating the species (as in Figure 1A), omitting *K.*

siamensis. The calculation proceeded in two steps. First, we fit a null model in which all

branches of the tree had the same protein evolutionary rate (PAML codeml run with options

verbose = 1, seqtype = 1, clock = 0, model = 2, NSsites = 2, fix_omega = 1, omega = 1),

yielding a protein evolutionary rate and log-likelihood of the data under the model. Next, we re-

fit using an alternative model in which the branch leading to *K. marxianus* was the foreground

(PAML codeml run with options verbose = 1, seqtype = 1, clock = 0, model = 2, NSsites = 2,

fix_omega = 0). Outputs were the codon sites assigned to the class with an excess of amino

acid changes in *K. marxianus* relative to silent changes, and their posterior probabilities under

Bayes Empirical Bayes inference (Yang et al. 2005); the codon sites assigned to classes with

relaxed and purifying selection, respectively, along the background lineages, and their posterior

probabilities; and the log-likelihood of the data under the model. The *p*-value assessing the

improvement in fit to the data of the alternative model relative to the null was calculated from a

likelihood ratio test (a χ^2 test on twice the difference of likelihoods with one degree of freedom).

p-values were corrected for multiple testing by the Benjamini/Hochberg method using the

multipletests function from statsmodels (Seabold and Perktold 2010). In Figure 5A we tabulated

all genes with adjusted *p* < 0.05 from this test that had more than 1% of codon sites assigned to

the class with an excess of amino acid changes in *K. marxianus* (proportion in class 2a > 0.01

in codeml). Separately, we repeated the above analyses using the branch leading to *K. lactis* as

the foreground. Conservation of sites called significant by PAML using *K. marxianus* as the

foreground were calculated using Biopython (Cock et al. 2009) to count the conservation of

each site among our sequenced *K. marxianus* strains (Table S1 and see Section 7.3).

Saccharomyces orthologs were taken from (Sorrells et al. 2018).

7.3. Resequencing and analysis of K. marxianus *and* K. lactis *population genomes*

For resequencing, we cultured, in 10 mL of liquid YPD, each of five *K. marxianus* strains of non-

domesticated provenance beside Km31, and two non-domesticated *K. lactis* strains beside

yLB72 (Table S1), and we isolated DNA from each with the Quick-DNA™ Fungal/Bacterial

Miniprep Kit (Zymo Research). Genomic DNA sequencing libraries were made and sequenced

on an Illumina NovaSeq 6000 by the UC Berkeley Vincent J. Coates Genomic Sequencing

Laboratory. We used Minimap2(Li 2018) to map the *K. marxianus* reads to our *K. marxianus*

Km31 reference genome (see Section 5.3). Separately, we mapped the reads from each of 41

K. lactis strains previously sequenced (Friedrich et al. 2023), and those from our two sequenced

K. lactis strains, to our *K. lactis* yLB72 reference genome. For each strain we sorted mappings

using the SAMtools fixmate and markdup functions, and we called single-nucleotide

polymorphisms and generated a pseudogenome as in Section 5.3, except the variant filtering

included only a quality score (QUAL) threshold of above 20. Called variants are provided in

Supplementary Data 2. For each pair of orthologous genes between *K. marxianus* and *K. lactis*

from section 5.3, we aligned the nucleotide sequences of the coding regions from all strains of

the two species using MUSCLE (Edgar 2004). We then used this alignment and its associated

nucleotide sequences as input into PAL2NAL (Suyama et al. 2006) using the -nogap flag to

remove gaps to generate a codon alignment.

For Figure 5D and 5E and Table S6 and S7, we carried out per-gene McDonald-Kreitman

analyses (McDonald and Kreitman 1991) on the PAL2NAL alignments using the mktest function

from the CodonAlignment class from Biopython. *p*-values were corrected for multiple testing by

the Benjamini/Hochberg method using the multipletests function from statsmodels (Seabold and

Perktold 2010). For a given gene, this approach tabulates the rates of non-synonymous

polymorphism within species (P_n), synonymous polymorphism (P_s), non-synonymous

divergence between species (D_n), and synonymous divergence (D_s), and the neutrality index NI

$= (P_n/P_s)/(D_n/D_s)$. Because low counts of amino acid polymorphism can drive significance in the

McDonald-Kreitman framework, we developed a filtering scheme in which we tabulated the

average a and standard deviation σ of P_n/P_s across all genes for which we had sequence data.

We only retained for Figure 5D those genes with $P_n/P_s > a - 1.5\sigma$.

For genomic analyses of P_n/P_s and D_n/D_s in the McDonald-Kreitman framework in Figure 5E and

Table S7, we used Gene Ontology term assignments and resampling (using 10,000 resamples)

as in Section 5.4, except that we only included Gene Ontology terms containing 10 genes or

more after filtering, and for P_n/P_s we tested for enrichment of low values of the metric across a

given term. Genomic analyses on NI proceeded as above, except on the filtered dataset of

genes with $P_n/P_s > a - 1.5\sigma$ as explained above and testing for enrichment of low values of the

metric across the term. *p*-values were corrected for multiple testing by the Benjamini/Hochberg

method using the multipletests function from statsmodels (Seabold and Perktold 2010).

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Data Availability

635   RNA-seq reads and normalized read counts will be made available in the Gene Expression

636   Omnibus, and genome sequencing reads in the Sequence Read Archive, upon publication.

Figure captions

Figure 1. *K. marxianus* shows a unique heat and chemical stress resistance phenotype.
A, Phylogenetic tree of the *Kluyveromyces* genus with marine and terrestrial species delineated (Am-In et al. 2008; Varela, Puricelli, Ortiz-Merino, et al. 2019). Branch lengths are not to scale. **B,** Each panel reports the results of a growth profiling experiment across the genus in the indicated condition. In a given panel, each trace reports a growth timecourse for the indicated species, with the solid line displaying the mean across technical replicates and the faint outline showing the standard error. OD, optical density.

Figure 2. The unique heat tolerance of *K. marxianus* only manifests during its growing stage. A, Experimental workflow created with BioRender.com. From left, inoculum from a starting liquid culture is introduced into fresh liquid medium for regrowth at 28°C and then sampled at nutrient exhaustion (Stationary) or in log phase (Exponential). The liquid sample is subjected to a heat shock of 47°C for 30 minutes and then spotted as serial dilutions onto solid medium. Viable cells after heat shock grow into colonies after incubation at 28°C. **B-C,** Each row reports viability after heat shock of a sample of the indicated species in the indicated liquid growth phase. See Figure S3 for unshocked controls.

Figure 3. Overall regulatory volatility by *K. lactis* relative to *K. marxianus* and stress-evoked repression of glycolysis in both species. A, Shown are results of principal component (PC) analysis of *K. lactis* and *K. marxianus* transcriptomes across treatments. Each point reports the values of the top two PCs for one replicate transcriptome of the indicated species subjected to the indicated stress treatment or a 28°C unstressed control. In a given axis label, the value in parentheses reports the proportion of variance across the set of

transcriptomes explained by the indicated PC. **B,C,** Each panel reports the effect of stress on

expression of genes of the indicated Gene Ontology term in *K. marxianus* and *K. lactis*. In a

given panel, each cell reports, for the indicated gene, the $\log_2$ fold-change between expression

levels in the indicated stress treatment and an unstressed control.

Figure 4. Gene family expansions in *K. marxianus*. A, The color of each cell reports the

change in copy number, in the indicated genome of a *Kluyveromyces* or related species relative

to its parent node in the phylogeny, of a gene family (orthogroup, OG) with significant expansion

or contraction in the *K. marxianus* genome as inferred with CAFE 5. Results for all analyzed

orthogroups are reported in Table S4. **B,** Each panel reports the positions of genes in the *K.*

marxianus genome in one indicated gene family from **A** with annotations as major facilitator

superfamily (MFS) transporters. In a given panel, each row represents a chromosome in *K.*

marxianus, and each arrow reports a gene of the respective family and its transcriptional

direction (not to scale).

Figure 5. Phylogenetic analysis reveals accelerated amino acid evolution at hundreds of

genes in *K. marxianus*. A, Each bar shows the number of genes with significant evidence for

positive selection on the indicated lineage (adjusted $p > 0.05$) in a phylogenetic PAML branch-

site test. **B,** The set of bars of each color reports the distribution, across the genes from **A**, of

the number of sites per gene inferred by PAML to be targets of positive selection on the

indicated lineage. **C,** Amino acid alignment of an example region of a gene, KLLA0B02607g (*S.*

cerevisiae ortholog FAT3/*YKL187C*), with evidence for positive selection from the PAML test in

A and from the McDonald-Kreitman test of **D**. Highlighted columns show inferred targets of

positive selection from the PAML test in **B**. Highlighting indicates the amino acid's BLOSUM 62

matrix score with respect to the consensus amino acid. **D,** Each row reports a gene with

evidence for adaptive protein variation between *K. marxianus* and *K. lactis* according to the

McDonald-Kreitman test (adjusted $p < 0.05$). NI, neutrality index. Genes highlighted in green

also showed significant evidence for positive selection on the branch leading to *K. marxianus* in

phylogenetic tests (PAML branch-site $p < 0.05$; Table S5). Not shown here are genes with *p*-

value < 0.05 as a product of exceptionally high or low amino acid polymorphism; see Methods

and Table S5. **E**, The *x*-axis reports mean NI across a group of genes. The blue trace reports

results from a resampling test for enrichment of low NI across the genes of the transmembrane

transport Gene Ontology term: the *y*-axis reports the frequency, among random sets of genes,

of the mean NI given on the *x*, and the true mean NI of genes of the transmembrane transport

term is indicated by the red dashed line, corresponding to corrected empirical $p = 0.0132$ (Table

S5).

Supplementary figure captions

Figure S1. Stresses in which *K. marxianus* does not exhibit unique stress tolerance

relative to the rest of the genus. Data and symbols are as in Figure 1B of the main text except

that each panel reports growth measurements in the indicated stress.

Figure S2. *K. marxianus* strains mirror each other in heat tolerance. Data and symbols are

as in Figure 1B of the main text except that in a given panel, each trace reports growth

measurements of the indicated wild *K. marxianus* strain.

Figure S3. No detectable difference between *Kluyveromyces* species with respect to cell

viability from unstressed liquid cultures. Data and symbols are as in Figure 2 of the main

text except that liquid cultures were not subjected to heat shock before plating.

Figure S4. Genetic or culture-based block of respiration compromises stress tolerance in

***K. marxianus* and *K. lactis*. A, B,** Each panel reports cell viability assays after culture in liquid

medium with stress, or an unstressed control, of *K. marxianus* or *K. lactis* in normoxia or

hypoxia. In a given panel, each row shows serial dilutions of the liquid culture spotted onto solid

medium and incubated without stress at 28°C.

Figure S5. Stress-evoked changes in *K. lactis*, but not *K. marxianus,* dominate

transcriptome comparisons between the two. A, Shown are results from the principal

component (PC) analysis of *K. lactis* and *K. marxianus* transcriptomes across treatments in

Figure 3A of the main text. Each panel reports expression across genotypes and treatments of

the three genes contributing most to the indicated PC. In a given panel, each bar reports

expression, as a mean across replicates, of the indicated gene in the indicated species and

treatment. The full set of gene weights is listed in Supplementary Table 2. **B, C**, Data are as in

Figure 3B,C of the main text except that genes of the indicated Gene Ontology term are

reported.

Supplementary table captions

Table S1. Strains used in this work. List of strains used for phenotypic assays and population analyses. All species shown are haploid.

Table S2. Normalized RNA-seq read counts, major principal component weights, and stress effects. Each row reports RNA-seq results and analysis for a pair of orthologous *K. lactis* and *K. marxianus* genes. Columns B and C report respective weighting of the gene in the first and second components of a principal component analysis. In columns D-S, each column reports normalized RNA-seq read counts as an average across replicates from cultures of *K. lactis* or *K. marxianus* in unstressed control conditions (LC and MC, respectively) or in stress treatments (LT or MT, respectively), where stresses are 39°C, ethanol (ET), MMS, or caffeine (CAF). The last eight columns report the effect of stress on expression; each reports the average across replicates of $\log_2$(treatment/control) for each species in the indicated treatment.

Table S3. Gene Ontology terms with stress-responsive expression. Each tab reports the results of tests of Gene Ontology terms for enrichment of expression change between the indicated stress and unstressed controls in the indicated species. In a given tab, each row reports results for the indicated Gene Ontology term: the average, across genes and replicates, of $\log_2$(treatment/control), and the multiple testing-corrected significance of the enrichment for high magnitude of this value relative to a resampling null.

Table S4. Gene family analysis for expansion/contraction analysis. Tab 1 reports the results from the CAFE analysis for gene family copy number change. Column B denotes the *p*-value from CAFE for a change in copy number in *K. marxianus* of the orthogroup indicated in

Column A. Column C shows the *p*-value after adjustment for multiple testing. Tab 2 shows the

genes called in each orthogroup for all species analyzed in CAFE.

Table S5. Phylogenetic tests for evidence of positive selection. In a given row in Tab 1,

columns C-I report the results of PAML branch-site tests package for positive selection in *K.*

marxianus on the indicated orthogroup (see Tab 2 for full orthogroup membership). Columns C-

D report the significance of the improvement in fit to an alternative model stipulating positive

selection relative to a null without positive selection, as a nominal *p*-value and after adjustment

for multiple testing. Columns E and G report the number and proportion, respectively, of sites

inferred to be under positive selection in the alternative model. Column F reports the average

conservation of the latter across *K. marxianus* population isolates from Supplementary Data 2.

Columns H and I report inferred protein evolutionary rates for the branch leading to *K.*

marxianus (foreground) and for all other branches of the tree (background), respectively.

Columns J-O report the analogous results from tests for positive selection in *K. lactis*. Columns

P-Q list the gene name and function for the *S. cerevisiae* ortholog. Green highlighted cells

indicate cases of *p* < 0.05. Tab 2 shows the genes in each orthogroup analyzed in PAML, with

each row reporting the genes of the indicated species representing the members of one inferred

orthogroup.

Table S6. McDonald-Kreitman tests for evidence of positive selection. Each row reports

results from a McDonald-Kreitman test on the indicated pair of orthologous genes between *K.*

lactis and *K. marxianus*. D_s, number of divergent synonymous changes between species; D_n,

number of divergent non-synonymous changes between species; P_s, number of polymorphic

synonymous changes within species; P_n, number of polymorphic non-synonymous changes

within species; NI, neutrality index (see Methods); *p* and *p*-value, nominal and multiple testing-

corrected McDonald-Kreitman significances, respectively. Columns L-M list the gene name and

function for the *S. cerevisiae* ortholog. Green highlighted cells indicate cases of $p < 0.05$.

Table S7. Gene Ontology terms with evidence for positive or purifying selection. Each tab

reports the results of tests of Gene Ontology terms for enrichment of extreme values of one

sequence-based metric of selection. These tests used the gene-based variation data from Table

S5 and were, respectively, for low values of the neutrality index, NI, indicative of positive

selection; for low values of P_s/P_n, indicative of purifying selection; and for high values of the

protein evolutionary rate D_s/D_n, indicative of positive selection. In a given tab, each row reports

results for the indicated Gene Ontology term: the average across genes of the indicated metric,

and the nominal and the multiple testing-corrected significance of the enrichment for extreme

magnitude of this value relative to a resampling null (p-value and p_adj, respectively).

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1041

Figure 1

A

B

Figure 2

A

B

Stationary

K. aestuarii	
K. dobzhanskii	
K. lactis	
K. marxianus	
K. nonfermentans	
K. siamensis	
K. wickerhamii	

C

Exponential

K. aestuarii	
K. dobzhanskii	
K. lactis	
K. marxianus	
K. nonfermentans	
K. siamensis	
K. wickerhamii	

Figure 3

A

B Nucleolus - GO:0005730

RCL1/KLLA0C05984g
MAK16/KLLA0A04037g
MRT4/KLLA0B13563g
NOP7/KLLA0D15268g
NOG1/KLLA0E24135g
NIP7/KLLA0B02266g
CGR1/KLLA0E10781g
RPC11/KLLA0F18568g
FYV7/KLLA0D05445g
NSA2/KLLA0E13619g
NSA1/KLLA0F10593g
SLX9/KLLA0C15565g
NOG2/KLLA0C18843g
UTP15/KLLA0F13772g
NHP2/KLLA0C15345g
SNU13/KLLA0E22529g
GAR1/KLLA0F21604g
RPB9/KLLA0D03234g
POL5/KLLA0D00792g

Km 39°C
Kl 39°C
Km caffeine
Kl caffeine
Km ethanol
Kl ethanol
Km MMS
Kl MMS

log₂(Treatment/control)

C Glycolysis - GO:0006096

TDH1-2/KLLA0F09141g
GLK1/KLLA0C01155g
HXK1/KLLA0D11352g
GPM2/KLLA0C04081g
PFK1/KLLA0A05544g
PFK2/KLLA0F06248g
PGK1/KLLA0A11011g
TPI1/KLLA0F18832g
ENO1/KLLA0A09185g
FBA1/KLLA0E07569g
GPM1/KLLA0E11859g
PGI1/KLLA0E23519g
CDC19/KLLA0F23397g
TDH3/KLLA0A11858g

Km 39°C
Kl 39°C
Km caffeine
Kl caffeine
Km ethanol
Kl ethanol
Km MMS
Kl MMS

Figure 4

A

Figure 5

K. lactis, K. marxianus genes	NI	Saccharomyces gene	Standard Name	Function
KLLA0E09329g, KM020H00950	0.19	YOR020W-A	MCO10	Subunit l of the mitochondrial F1F0 ATP synthase
KLLA0E07569g, KM020D04120	0.26	YKL060C	FBA1	Fructose 1,6-bisphosphate aldolase
KLLA0D10780g, KM020E01340	0.32	YJR118C	ILM1	Unknown function, possible mitochondrial involvement
KLLA0C12133g, KM020H02920	0.39	YNL284C	MRPL10	Mitochondrial ribosomal protein
KLLA0A10175g, KM020A03410	0.47	YMR313C	TGL3	Triacylglycerol lipase and LPE acyltransferase
KLLA0F19162g, KM020D01640	0.49	YDR111C	ALT2	Catalytically inactive alanine transaminase
KLLA0B04994g, KM020F01720	0.50	YLR443W	ECM7	Putative integral membrane protein, calcium uptake
KLLA0F00528g, KM020F00240	0.51	YAL061W	BDH2	Putative medium-chain alcohol dehydrogenase
KLLA0B02607g, KM020D05580	0.59	YKL187C	FAT3	Fatty acid uptake
KLLA0D16236g, KM020F04880	0.63	YGR130C	N/A	Component of the eisosome
KLLA0F26048g, KM020E04040	0.68	YMR066W	SOV1	Mitochondrial protein of unknown function

Figure S1

Figure S2

A

28°C

B

42°C

K. marxianus UCD61-293
K. marxianus UCD55-67
K. marxianus UCD40-39
K. marxianus UCD55-69
K. marxianus UCD05-822

Figure S3

Figure S4

A *K. marxianus*

B *K. lactis*

Aerobic   Hypoxic

Control
39°C
25mM Caffeine
6.5% Ethanol
0.03% MMS

Figure S5

A

B

C