Complex Genetic Interactions in a Quantitative Trait Locus

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Whether in natural populations or between two unrelated members of a species, most phenotypic variation is quantitative. To analyze such quantitative traits, one must first map the underlying quantitative trait loci (QTLs). Next, and far more difficult, one must identify the quantitative trait genes (QTGs), characterize QTG interactions, and identify the phenotypically relevant polymorphisms to determine how QTGs contribute to phenotype. In this work, we analyzed three Saccharomyces cerevisiae high-temperature growth (Htg) QTGs: MKT1, END3, and RHO2. We observed a high level of genetic interactions among QTGs and strain background. Interestingly, while the MKT1 and END3 coding polymorphisms contribute to phenotype, it is the RHO2 3’UTR polymorphisms that are phenotypically relevant. Reciprocal hemizygosity analysis of the Htg QTGs in hybrids between S288c and ten unrelated S. cerevisiae strains reveals that the contributions of the Htg QTGs are not conserved in nine other hybrids, which has implications for QTG identification by marker-trait association. Our findings demonstrate the variety and complexity of QTG contributions to phenotype, the impact of genetic background, and the value of quantitative genetic studies in S. cerevisiae.

Introduction

Identification of quantitative trait genes (QTGs) has been hampered by multiple factors, including variable quantitative trait locus (QTL) contributions, complex architectures [1,2], epistasis [3,4], gene-environment interactions, and genetic heterogeneity [5–7]. With the availability of whole genome sequences, several strategies have been proposed to facilitate the identification of QTGs [7–12], including marker-trait association, systematic functional screening, and candidate gene prediction by expression profiling or phenotype screening [13]. However, the major gap in QTL analysis is the lack of detailed molecular information for any one quantitative trait regarding QTL structure, genetic interactions, and the contributions (or relevance) of QTGs identified in one genetic background or population to other genetic backgrounds or populations of the same species.

We previously mapped a Saccharomyces cerevisiae high-temperature growth (Htg) QTL and identified MKT1, END3, and RHO2 as Htg QTGs [1]. The QTL complexity was surprising not only because there were three QTGs in the 32-kb interval, but also because the contributing alleles derived from both the Htg+ YM145 (alleles designated as MKT1–145 and RHO2–145) and Htg– S288c (END3–288) backgrounds. In the YM145, S288c, and 11 other Htg– genetic backgrounds, there were multiple coding and noncoding QTG polymorphisms; however, by marker-trait association the phenotypically relevant polymorphisms were not identifiable [1].

In this study, we assessed genes encoding genetic or physical interactors of MKT1, RHO2, and END3 as novel QTGs and found that none of these genes are Htg QTGs; we also characterized the three QTGs in isolation and in different combinations using deletion and reciprocal hemizygosity analysis (RHA; see Figure 1) [1]. We performed expression and homologous allele replacement analysis to identify phenotypically relevant polymorphisms and used RHA to determine allele contributions to Htg in ten different genetic backgrounds. Our study sheds light on the complexity underlying quantitative traits and emphasizes the utility of different approaches to dissect and analyze quantitative traits.

Results

Known Genetic and Physical Interactors of the Htg QTGs Are Not Htg QTGs Themselves

One potential strategy to identify novel QTGs is to test genes that interact genetically (e.g., synthetic lethality) or physically (e.g., in complexes) with known QTGs. No genetic interactions have been reported for mkt1 or rho2 mutations [14,15]. However, end3 has been shown to be synthetically lethal with apm3 [16], arp2 [17], and pan1 [18] mutations and has a synthetic slow-growth phenotype with rvs161A and rvs167A deletions [19]. End3p also physically interacts with Pbp1p [20] and, in addition, Mkt1p physically interacts with Pbp1p [21]. Therefore, we performed RHA in YJM145/S288c hybrids to determine if these genes, the gene products of which interact genetically or physically with the gene products of known Htg QTGs, are Htg QTGs themselves.

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Abbreviations: Htg, high-temperature growth; QTG, quantitative trait gene; QTL, quantitative trait locus; QTN, quantitative trait nucleotide; RHA, reciprocal hemizygosity analysis; SNP, single-nucleotide polymorphism; UTR, untranslated region

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Synopsis

Most of the differences in phenotype between unrelated members of a species are polygenic in nature. Because of their ubiquity and importance, these polygenic (or quantitative) traits have been intensively studied, and a variety of techniques have been proposed to identify and characterize quantitative trait genes (QTGs). Indeed, the main application of the recently published human HapMap project is to identify the genes responsible for diseases that are quantitative in nature. Using a well-defined Saccharomyces cerevisiae quantitative trait locus containing three QTGs (MKT1, END3, and RHO2), the authors used deletions to analyze the contributions of each gene to phenotype, singly and in combination, and found a variety of interactions. Expression analysis showed no difference in steady-state mRNA levels between alleles of the three genes. Homologous allele replacement identified the phenotypically relevant differences between alleles of each gene, which were single coding polymorphisms for two genes (MKT1 and END3) and the 3′ untranslated region of one gene (RH02). Finally, analysis of multiple genetic backgrounds showed that the phenotypes conferred by these genetic variants were not conserved. The results show that the techniques proposed to identify QTGs, such as expression analysis and marker-trait association, have profound limitations, and that unbiased genome-wide approaches are needed to dissect quantitative traits. The results also demonstrate the complexity of the genetic interactions that affect quantitative traits and the value of the S. cerevisiae system in studying these traits.

However, RHA showed that none of these interacting genes make allele-specific contributions to the Htg phenotype; that is, APA13, ARP2, PAN1, RVS161, RVS167, and PBP1 are not Htg QTGs.

Deletion Analysis of Htg QTGs

To determine the deletion phenotypes of Htg QTGs in the S288c and YJM145 backgrounds, we constructed single (qtg1Δ/qtg1Δ) and double (qtg1Δ/qtg1Δ qtg2Δ/qtg2Δ) diploid deletion strains and tested their Htg phenotypes. Consistent with previous work [21,22], end3A S288c and YJM145 background strains were temperature-sensitive at 37 °C. Therefore, End3p is required for growth at even modestly elevated temperatures in both backgrounds.

In the S288c background at 39 °C, the highest temperature tested in this Htg background, rho2Δ strains grew better than wild-type strains. However, in the Htg S288c background, rho2Δ strains grew less than wild type at 40 °C and 41 °C (Table 1). In the S288c background, mkt1A and wild-type strains grew equivalently at 39 °C. However, in the YJM145 background, mkt1A strains grew less than wild type at 40 °C, while at 41 °C mkt1A strains did not grow. Therefore, while end3A mutants in both parental backgrounds were unable to grow at 37 °C, the rho2A and mkt1A phenotypes were different in the two genetic backgrounds, which suggests that phenotype screening of null mutants may have limited predictive value as a QTG identification strategy.

In both the S288c and YJM145 backgrounds, we constructed double mutants to test for synthetic interactions between individual Htg QTGs. No synthetic lethal interactions were noted for any pair of deletions or for deletion of all three Htg QTGs in either background. All end3Δ-containing strains (end3A mkt1Δ, end3A rho2Δ, and end3A mkt1Δ rho2Δ) grew similarly to end3Δ strains in both backgrounds. However, in contrast to rho2Δ and mkt1A strains, mkt1Δ rho2Δ strains in the S288c background grew less than wild type at 39 °C, and thus exhibit a synthetic Htg phenotype (Table 1). When tested in the YJM145 background at 40 °C, however, the growth defect of mkt1Δ rho2Δ strains was the sum of the mkt1Δ and rho2Δ growth defects, consistent with Mkt1-145p and Rho2-145p contributing additively to Htg (Table 1).

Genetic Interactions between Htg QTG Alleles

As the first step in assessing the genetic interactions between Htg QTG alleles, we determined the Htg phenotypes of a heteroallelic S288c/S288c hybrid versus deletion heterozygotes; for example, END3–145/RHO2–288 versus END3–145/RHO2–288 and END3–145/RHO2–288 versus END3–145/RHO2–288. Consistent with the Htg plus alleles of MKT1, END3, and RHO2 being dominant, deletion heterozygotes containing only the Htg plus QTG alleles had Htg phenotypes similar to a heteroallelic hybrid containing both Htg plus and Htg minus alleles (Table 2).

Having determined the Htg phenotypes of a heteroallelic hybrid versus all single-deletion heterozygotes, we tested for interactions between Htg QTG alleles by determining the relative Htg phenotypes of a heteroallelic hybrid versus double-deletion heterozygotes. In competitions of a heteroallelic hybrid versus hybrids with the Htg minus alleles of RHO2 and END3 deleted (RHO2–145/Δ END3–288), there was no...
difference between the hybrids. In contrast, in competitions of the heteroallelic hybrid versus hybrids with the Htg\textsuperscript{+} alleles of RH02 and END3 deleted (\textit{A/RH02–288 END3–145/A}), the deletion heterozygotes were Htg\textsuperscript{+} relative to the heteroallelic hybrid. However, the approximately two-fold growth differential in these competitions was much less than that observed when single Htg\textsuperscript{+} alleles of RH02 and END3 were deleted (Table 2), suggesting a complex genetic interaction between alleles of RH02 and END3.

In competitions of a heteroallelic hybrid versus hybrids with Htg\textsuperscript{−} alleles of \textit{MKT1} and END3 deleted (\textit{A/MKT1–288 END3–145/A}), the deletion heterozygotes had Htg phenotypes significantly less than the heteroallelic hybrid. The Htg phenotypic differential in these competitions was approximately equal to the sum of the differences observed in the single-deletion heterozygote analyses (Table 2), indicating \textit{MKT1–145} and \textit{END3–288} additivity. Similarly, in competitions of a heteroallelic hybrid versus hybrids with Htg\textsuperscript{−} alleles of \textit{MKT1} and RH02 deleted (\textit{A/MKT1–288 A/RH02–288}), the deletion heterozygotes had Htg phenotypes significantly less than the heteroallelic hybrid, consistent with the earlier observed \textit{MKT1–145} and \textit{RH02–145} additivity.

### Table 1. Phenotypes of Htg QTG Deletion Strains in the Parental Backgrounds

| Genotype          | Fold Growth Differences (+/+ versus A/A)* |
|-------------------|------------------------------------------|
|                   | S288c Background, 39 °C | YJM145 Background, 40 °C | YJM145 Background, 41 °C |
| rho2A/rho2A       | 0.51**                                  | 1.53*                      | 2.62**                     |
| mkt1.1/mkt1.1     | 0.85                                    | 3.92***                    | No growth\textsuperscript{D} |
| mkt1.1/mkt1.1     | 7.12***                                 | 6.64***                    | No growth\textsuperscript{D} |

*Calculated as (mean cfu\textsubscript{+/+}/mean cfu\textsubscript{A/A}) between isogenic \textit{+/+} and A/A strains.

** Significant at \(p < 0.001\).

* Significant at \(p < 0.05\).

\(\ast \) Significant at \(p < 0.01\).

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According to the results, the deletion heterozygotes had Htg phenotypes significantly less than the heteroallelic hybrid. The phenotypic differential in these competitions was approximately equal to the sum of the differences observed in the single-deletion heterozygote analyses (Table 2), indicating \textit{MKT1–145} and \textit{END3–288} additivity. Similarly, in competitions of a heteroallelic hybrid versus hybrids with Htg\textsuperscript{−} alleles of \textit{MKT1} and RH02 deleted (\textit{A/MKT1–288 A/RH02–288}), the deletion heterozygotes had Htg phenotypes significantly less than the heteroallelic hybrid, consistent with the earlier observed \textit{MKT1–145} and \textit{RH02–145} additivity.

### Table 2. Growth Difference at 41 °C between a Heteroallelic Hybrid and Hybrids Heterozygous for Deletions of One or Two Htg QTGs

| Genotype               | Fold Difference* |
|------------------------|------------------|
| \textit{MKT1–145/A}    | 2.02             |
| \textit{A/MKT1–288}    | 7.46***          |
| \textit{RH02–145/A}    | 0.98             |
| \textit{A/RH02–288}    | 3.71***          |
| \textit{END3–145/A}    | 8.24***          |
| \textit{A/END3–288}    | 1.31             |
| \textit{RH02–145/A}    | 1.34             |
| \textit{A/RH02–288}    | 1.98**           |
| \textit{MKT1–145/A}    | 2.09**           |
| \textit{A/MKT1–288}    | 13.41***         |
| \textit{MKT1–145/A}    | 12.34            |
| \textit{A/MKT1–288}    | 14.91***         |

*Calculated as (mean cfu\textsubscript{+/+}/mean cfu\textsubscript{A/A}) between the heteroallelic YJM145/S288c hybrid XHS123 (\textit{A/MKT1–288 RH02–288 END3–145/A}) and isogenic strains heterozygous for deletions of one or two Htg QTGs. Also see Table S1.

** Significant at \(p < 0.01\).

*** Significant at \(p < 0.001\).

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### Expression Analysis of the Htg QTGs

To further characterize the Htg QTGs and gain insight into the mechanism of the quantitative trait and the location(s) of the phenotypically relevant polymorphism(s), we carefully examined the steady state mRNA levels of the three Htg QTGs as a function of both temperature and allele. Since isogenic reciprocal hemizygotes contain only one allele of a QTG, in a comparison between reciprocally hemizygous strains, any allele-specific differences in steady state mRNA levels must be due to \textit{cis}-acting polymorphisms. Therefore, we performed Northern analysis of two pairs of reciprocal hemizygotes for each QTG. Temperature (25 °C versus 37 °C) had no effect on RH02 and END3 steady state mRNA levels. However, MKT1 steady state mRNA levels increased in an allele-nonspecific manner at 37 °C (37 °C/25 °C = 2.0). No allele-specific fold differences larger than two-fold were observed in steady state mRNA at either temperature for any of the genes.

### Homologous Allele Replacement of the Htg QTGs

As a first test of the hypothesis that the phenotypically relevant differences between Htg\textsuperscript{+} and Htg\textsuperscript{−} QTG alleles were contained within the genes, and most likely from expression results within the coding regions, we ectopically integrated alleles of RH02 and alleles of MKT1 (separately) at the \textit{HO} locus. However, we found that the resulting strains had insert-specific, Htg QTG allele-nonspecific, dominant-negative growth defects (unpublished data).

Since these ectopic integrations were problematic, we then constructed isogenic homologous allele replacement strains that differed (see Table S2) solely with respect to the S288c versus YJM145 nonsynonymous coding region polymorphisms and one 5' and one 3' UTR polymorphism in \textit{MKT1} (a−104g D30G K453R a+54t), or nonsynonymous coding polymorphisms in \textit{END3} (S258N D268N) or RH02 (F91C) (see Table S3). We then compared the Htg phenotypes of isogenic strains that differed solely in these \textit{MKT1}, \textit{END3}, or \textit{RH02} polymorphisms.

For isogenic S288c, YJM145 and YJM145/S288c hybrid background strains with different RH02 alleles, there were no statistically significant differences in Htg phenotype (Table 3). Therefore, the single nonsynonymous RH02 coding polymorphism (C91F) is phenotypically neutral. Since there are no RH02 5'UTR polymorphisms, this suggested that RH02 3'UTR polymorphisms contribute to phenotype (see Table S3 for polymorphisms).
To test the RHO2 3’UTR hypothesis, we replaced the 3’UTR of RHO2–288 with the 3’UTR of RHO2–145; that is, we constructed RHO2–288 3’UTR–145 strains. We then performed Htg competitions between heteroallelic hybrids (RHO2–145 3’UTR–145/RHO2–288 3’UTR–288) versus hybrids with the S288c-derived RHO2 3’UTR (ΔRHO2–288 3’UTR–288) versus hybrids with the YJM145-derived RHO2 3’UTR (ΔRHO2–288 3’UTR–145). Consistent with our RHA results, the ΔRHO2–288 3’UTR–145 strains grew as well as the heteroallelic hybrids, while the growth difference between ΔRHO2–288 3’UTR–288 and the heteroallelic hybrid was similar to the growth difference between ΔRHO2–288 3’UTR–288 and ΔRHO2–288 3’UTR–145 strains (Table 4). Therefore, the RHO2 3’UTR is primarily and possibly solely responsible for the allele-specific contribution of RHO2 to the Htg phenotype.

For both isogenic S288c and YJM145 background strains with different END3 coding alleles, there were no significant differences in Htg phenotype (Table 3). However, consistent with our previous results [1], Htg END3–288 allele-containing YJM145/S288c hybrid background strains had highly significant Htg advantages over Htg END3–145 allele-containing strains.

In the S288c background, there was no significant difference in growth between strains with different MKT1 alleles (Tables 3 and S2). However, in the YJM145 background, Htg MKT1–145 allele-containing strains had significantly higher Htg advantages over Htg MKT1–288 allele-containing strains. Similarly, in the YJM145/S288c hybrid background, Htg MKT1–145 allele-containing strains had highly significant Htg advantages over Htg MKT1–288 allele-containing strains. Therefore, these END3 coding (S258N D268N) and MKT1 coding (a–D30G K453R a+54t) polymorphisms are responsible for much of the END3 and MKT1 allele-specific contributions to the Htg phenotype, and their effects are highly influenced by genetic background.

Identification of the MKT1 and END3 Quantitative Trait Nucleotide

To determine which of the coding single-nucleotide polymorphisms (SNPs) in the two major Htg QTGs, MKT1 and END3, are relevant quantitative trait nucleotides (QTNs) for the Htg phenotype, we used site-directed mutagenesis to replace the two coding SNPs in MKT1 and END3 individually. The isogenic strains were created such that they differed solely in the coding SNP under investigation. RHA competitions were done to determine the effect of each SNP replacement in each gene on the Htg phenotype.

For MKT1, RHA analysis of SNP replacement strains showed that the A1358G SNP (where the S288c base-nucleotide numbers from the start-YJM145 base), resulting in a conservative amino acid change from lysine (K) in S288c to arginine (R) in YJM145 (K453R), had no effect on Htg. However, the A899G SNP, resulting in a nonconservative amino acid change from aspartic acid (D) in S288c to glycine (G) in YJM145 (D30G), affected Htg with 89G (30G)-containing strains growing better at high temperature (Table 5).

For END3, RHA analysis of SNP replacement strains showed that the C802T SNP, resulting in an amino acid change from aspartic acid (D) in S288c to asparagine (N) in YJM145 (D30G), affected Htg with 89G (30G)-containing strains growing better at high temperature (Table 5).

Table 3. Growth Difference between Coding Allele Exchange Strains for Htg Genes in the S288c, YJM145, and YJM145/S288c Hybrid Backgrounds

| Htg QTG Genotypes | Fold Difference* |
|--------------------|------------------|
|                    | S288c Background a (39 °C) | YJM145 Background b (41 °C) | YJM145/S288c Background c (41 °C) |
| END3–288 versus END3–145 d | 1.20 | 0.83 | 3.93** |
| MKT1–145 versus MKT1–288 d | 0.55 | 182.54*** | 3.91** |
| RHO2–145 versus RHO2–288 d | 1.97 | 2.83 | 1.11 |

*Calculated as (mean cfuStrain1)/mean cfuStrain2. Also see Table S1.
** Significant at p < 0.01.
*** Significant at p < 0.001.
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Table 4. Growth Difference at 41 °C between YJM145/S288c Hybrids Containing YJM145- versus S288c-Derived RHO2 3’UTRs

| Genotype | Fold Difference* |
|----------|------------------|
| (RHO2–145 3’UTR–145/RHO2–288 3’UTR–288) versus (ΔRHO2–288 3’UTR–145) | 1.07 |
| (RHO2–145 3’UTR–145/RHO2–288 3’UTR–288) versus (ΔRHO2–288 3’UTR–288) | 3.14** |
| (ΔRHO2–288 3’UTR–288 versus ΔRHO2–288 3’UTR–145) | 2.94** |

*Calculated as (mean cfuStrain1)/mean cfuStrain2. Also see Table S1.
** Significant at p < 0.01.
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Table 5. Growth Difference at 41 °C between YJM145/S288c Hybrids Containing YJM145- versus S288c-Derived MKT1 and END3 Coding Polymorphisms

| Genotype | Fold Difference* |
|----------|------------------|
| MKT1 (30G) versus MKT1 (30D) | 12.9*** |
| END3 (2585) versus END3 (258N) | 36.6** |

*Calculated as (mean cfuStrain1)/mean cfuStrain2. Also see Table S1.
** Significant at p < 0.01.
*** Significant at p < 0.001.
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affected Htg with 773C (258S)-containing strains growing better at high temperature (Table 5).

Phenotypic Heterogeneity among the Htg QTGs in Diverse Genetic Backgrounds

Marker-trait association studies assess the association between a sequence variant and a trait. Marker-trait association, which may be affected by genetic background [2,3] and is one of the applications of the data coming from the HapMap project, aims to aid QTG discovery [23]. As a first step in determining the impact of genetic background on phenotype and on marker-trait association, we determined the Htg phenotype of the S288c background and of ten additional unrelated genetic backgrounds. The Htg phenotypes of the 11 unrelated genetic background strains varied over a greater than 104-fold range, demonstrating a high degree of Htg phenotypic heterogeneity, and there was no association between the specific MKT1, RHO2, and END3 polymorphisms shown to be relevant in YJM145/S288c and the Htg phenotype (Table 6).

To further assess the impact of genetic background on phenotype and on marker-trait association, we crossed an S288c background strain with strains from each of the ten unrelated Htg genetic backgrounds and determined the Htg phenotypes of the resulting hybrids. All hybrids, except YJM627/S288c, showed positive heterosis (Table 6); that is, the Htg phenotype of the hybrids was greater than either parent, consistent with a general benefit of heterozygosity at many loci. Therefore, although the S288c genome constitutes one-half of the genome content of all hybrids, the genetic composition of the entire genome affects the Htg phenotype. Again, there was no association between the specific MKT1, RHO2, and END3 polymorphisms shown to be relevant in YJM145/S288c and Htg phenotype.

As a final step in determining the impact of genetic background on phenotype and on marker-trait association, we performed RHA of the Htg QTGs in the ten hybrid genetic backgrounds (Table 6) [1]. RHA of MKT1, END3, and RHO2 in hybrids between S288c and the ten unrelated genetic backgrounds further demonstrated the complex impact of genetic background on QTG contributions to phenotype. For example, in MKT1, all of the non-S288c backgrounds have the same coding polymorphism (30G) that was found to be Htg in the YJM145/S288c hybrid. However, in the hybrids between S288c (30D) and these other genetic backgrounds, the 30G-containing MKT1 alleles present in all of the hybrids could be phenotypically Htg, neutral, or Htg and, conversely, the MKT1–288 (30D) allele present in all of the hybrids could be Htg, neutral, or Htg. There are similar cases for the END3 Htg coding polymorphism 258S. Our analysis shows therefore that QTGs identified in one background do not necessarily make the same contribution to phenotype in other genetic backgrounds, and that QTL architecture (that is, the

Table 6. Htg Phenotypes and RHA of Htg Genes in Hybrids between S288c and Other S. cerevisiae Genetic Backgrounds

| Genetic Background | Htg: YJM versus S288c | Htg: Hybrid versus S288c | MKT1 | END3 | RHO2 |
|--------------------|-----------------------|-------------------------|------|------|------|
|                    |                       |                         | 30 Htg Allele (Fold Difference) | 258 Htg Allele (Fold Difference) | +6 +37 +120 +262 +343 Htg Allele (Fold Difference) |
| S288c              |                       |                         | D – | S – | Δ g t g g – |
| YJM145             | +14.5                 | +8,320                  | G MKT1–145 (9.8) | N – | a c c a a RHO2–14 (4.0) |
| S288c              |                       |                         | D – | S – | Δ g t g g – |
| YJM42              | +323.6                | +17,827                 | G MKT1–421 (102.8) | S = | a c c g a RHO2–421 (17.5) |
| S288c              |                       |                         | D nt | S = | Δ g t g g – |
| YJM339             | +58.6                 | +351                    | G nt | S = | a c t g a |
| S288c              |                       |                         | D = | S = | Δ g t g g nt |
| YJM320             | +1.5 (NS)             | +877                    | G = | N = | Δ g t g g nt |
| S288c              |                       |                         | D MKT1–288 (8.0) | S = | Δ g t g g nt |
| YJM280             | +1.0 (NS)             | +2.0                    | G – | N = | a c c g a nt |
| S288c              |                       |                         | D = | N = | Δ g t g g = |
| YJM627             | +0.9 (NS)             | +1.0 (NS)               | G = | N = | Δ g t g g = |
| S288c              |                       |                         | D MKT1–288 (16.9) | S = | Δ g t g g = |
| YJM270             | +6.7                  |                         | G – | S = | Δ g t g g = |
| S288c              |                       |                         | D = | S = | Δ g t g g = |
| YJM326             | +1.42                 |                         | G MKT1–326 (18.0) | N = | Δ g t g g = |
| S288c              |                       |                         | D = | S = | Δ g t g g = |
| YJM1076            | +11.7                 |                         | G – | S = | Δ g t g g = |
| S288c              |                       |                         | D = | S = | Δ g t g g = |
| YJM1129            | +49.3                 |                         | G MKT1–1129 (43.2) | N = | Δ g t g g = |

Phenotypically relevant polymorphisms in the YJM145, S288c and YJM145/S288c genetic backgrounds (bold – Htg) as well as the corresponding polymorphisms in nine other genetic backgrounds. The locations of RHO2 3′UTR polymorphisms are given as (+) position from stop codon. The amino acid positions, with Htg polymorphisms indicated in bold, are given from start of protein sequence. For a complete list of MKT1, END3, and RHO2 polymorphisms in all genetic backgrounds, see Table S3.

*Fold Htg difference between the S288c background and each YJM background at 41 °C (+) indicates that a YJM background strain grows more at 41 °C than the S288c background. (−) denotes that a YJM background strain has a Htg phenotype at 41 °C with at least 200-fold less growth than the S288c background strain. All competitions were done in diploid strains. All the differences are significant (p < 0.01) unless noted as nonsignificant (NS).

*Fold Htg difference between the S288c background and each hybrid YJM/S288c background at 41 °C (+) indicates that a hybrid background strain grows more at 41 °C than the S288c background. All competitions were done in diploid strains. All the differences are significant (p < 0.01) unless noted as nonsignificant (NS).

*Htg allele (fold growth differences between isogenic reciprocal deletion heterozygotes at 41 °C as determined by RHA; allele designations correspond to strain/genetic background number, e.g., S288c, YJM145, etc. Whenever a YJM-derived allele is designated as being Htg, the S288c-derived allele is Htg and denoted as (−). (−) indicates that the reciprocal deletion heterozygotes had the same Htg phenotype; that is, the alleles did not differ in their contributions to the Htg phenotype. (nt) indicates not tested because the gene could not be deleted in the YJM genetic background.

See Table S1.

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identity and arrangement of QTGs) is therefore not conserved and instead varies with genetic background.

**Discussion**

Here we greatly extended our previous analysis [1] by determining the genetic interactions between the Htg QTGs and identifying the QTNs in the Htg genes. We determined the conservation of these QTNs in various strain backgrounds and tested various methods of QTG identification and analysis.

Our results show that not only is the architecture of this QTL complex, but that genetic interactions further add to the complexity. We found both additive interactions, such as between MKT1 and RHO2, and more complex genetic interactions, between END3 and RHO2. In addition, we showed that alleles have different effects in different genetic backgrounds. For example, END3–288 and END3–145 are phenotypically equivalent in the parental backgrounds, but differ phenotypically in the hybrid background. That is, alleles of genes from both parental backgrounds are needed to produce the END3–288 versus END3–145 phenotypic differential.

We find that none of the known interacting genes of the Htg QTGs are themselves Htg QTGs. We also find that none of the three Htg QTGs differ in their steady state mRNA levels. Indeed, our results show that RHO2 3'UTR polymorphisms that have no effect on steady state mRNA levels but may instead affect mRNA translation [24,25] or localization [26] contribute to Htg; as with QTGs containing coding QTNs, such QTGs are not identifiable by expression analysis. Since the above functional and expression techniques form the basis for candidate gene approaches to identify QTGs, our results demonstrate the need for unbiased genome-wide approaches to identify QTGs.

These three Htg QTGs have no obvious functional relationship, which may have implications for other quantitative traits. END3, which has a temperature-sensitive deletion phenotype [22] and a phenotypically relevant coding polymorphism, is involved in endocytosis. RHO2 encodes a nonessential small GTPase [27]; we found that no allele-specific differences exist in RHO2 steady state mRNA levels and that the RHO2 3'UTR is phenotypically relevant. MKT1 is involved in the maintenance of the M2 dsRNA satellite of the L-A virus; the S288c-allele is thought to be a loss-of-function mutation [28]. MKT1 also plays a role in the posttranscriptional regulation of HO by binding to the 3'UTR of HO mRNA, without affecting steady state mRNA levels [20]. We find that the MKT1 D30G polymorphism affects Htg and others have shown that this same MKT1 polymorphism also plays a role in sporulation efficiency [29]. Since none of the mkl1 mutant phenotypes appear to be related, this represents an intriguing case of pleiotropy. This mkl1 mutant pleiotropy may be related to the function of Mkt1p in binding to the 3'UTR of HO mRNA and controlling translation (without affecting steady state mRNA levels); in addition to HO, Mkt1p may regulate the translation of other mRNAs, such as RHO2.

Our analysis of Htg allele contributions in multiple genetic backgrounds demonstrates the impact of genetic background on allelic phenotype. Although our sample size was small, the lack of association of the sequence variants to the contribution of an allele has implications for marker-trait association. A variety of theoretical and experimental studies of QTLs in higher organisms [5,7,8,30] have suggested that epistasis and/or genetic background is a complicating factor in QTG identification; our results provide a graphic demonstration of these effects. To fully understand a phenotype, one must understand the background effects on phenotype, which requires the identification of the genetic variants in each background that are causative. One way to identify the causative genetic variants is to “homozygose” known QTGs, and then map novel QTLs; another is to backcross F1 segregants and map novel Htg QTLs. These strategies will assist in identifying additional Htg QTLs that contribute to phenotype in the presence and absence of the identified Htg QTGs.

Our results show the importance of yeast as a model system for QTL analysis and highlight important issues related to QTL identification and analysis. Our results and conclusions from quantitative studies in yeast show that the study of these phenotypes in higher organisms remains a difficult challenge. Improved understanding of quantitative traits in simpler eukaryotes such as yeast will facilitate the analysis of complex traits in higher eukaryotes.

**Materials and Methods**

**Yeast strains, growth media, and conditions.** Yeast strains were grown at 30 °C unless otherwise stated. The S. cerevisiae strains used in this work (Table S4) are derivatives of S288c (a laboratory genetic background, designated “S”), YJM145 (a clinically-derived genetic background, designated “YBN,” “YJM,” or “YHS” [31]), and others clinically and nonclinically derived genetic backgrounds (designated YHS, YRP, and YJM). To make hybrids (designated as “XHS”) of test and S288c background strains, spores of lys2 and lys5 diploid strains were allowed to mate on YPD plates that were then replica plated to SDC-LYS plates to select Lys+ diploid hybrids. YPD (yeast extract peptone and dextrose), YPEG (yeast extract peptone ethanol-glycerol), SD (synthetic dextrose minimal), SDC (synthetic dextrose complete) [32], and α-aminoacidate media [33] have been described previously. Filter-sterilized antibiotics (hygromycin 300 μg/ml, nourseothricin 100 μg/ml, G418 200 μg/ml) were added to YPD after autoclaving [34,35]. Methomyl (Chem Service, West Chester, Pennsylvania, United States) was added to YPEG medium (1 mM final concentration) that had cooled to about 50 °C; plates were poured immediately after mixing.

To induce sporulation, cultures were first grown overnight in liquid YPD at 30 °C, after which a 1-ml aliquot was removed, washed in sterile water twice, and resuspended in 3 ml of sterile 0.5% (w/v) potassium acetate medium [36]. Sporulating cultures were incubated at 30 °C for 48 h with rotation in a roller drum and examined microscopically for the presence of tetrads.

High-temperature growth competitions between two or three strains were performed using a technique similar to RHA, as previously described [1].

**PCR and yeast transformation.** The PCR to make gene deletion constructs containing dominant drug-resistance markers was done as described previously [35]. All PCR products were purified with Qiagen's PCR Purification kit (Qiagen, Valencia, California, United States) before introduction into yeast. DNA was introduced into yeast by the lithium acetate transformation protocol [37]. Homologous integration of dominant drug-resistance cassettes was verified by colony PCR [38]. In cases where colony PCR did not yield a product, genomic DNA was isolated [39] and PCR was performed using colony PCR conditions. The primers used in the study (IDT, Coralville, Iowa, United States) are listed in Table S5.

**Reciprocal hemizygosity analysis.** RHA, a technique for identifying allele-specific contributions to a phenotype, was performed as previously described (see Figure 1) [1]. The resulting data were used to calculate means and standard errors and to perform tests of significance [40].

**Reciprocal hemizygosity expression analysis.** To assay the expression levels of each allele of the Htg quantitative trait genes, we performed Northern analysis of isogenic reciprocal hemizygotes. For each Htg quantitative trait gene, a pair of reciprocal hemizygotes were grown (separately) at 25 °C and 37 °C in 50 ml of YPD with aeration to OD600 of 1.0. Total RNA was extracted as previously described, with the following modifications. After hot phenol-chloroform extraction, samples were frozen overnight at −80 °C and then centrifuged for 15 min at 13,000 rpm; the aqueous phase was processed as previously.
described. Cells were then washed and resuspended in 400 μl of extraction buffer (50 mM sodium acetate [pH 5.3], 1 mM EDTA) and 40 μl of SDS (10% w/v) followed by extraction with 440 μl of hot buffered phenol-chloroform (65 °C) for 30 min with vortexing every 5 min. After being frozen overnight at −80 °C, samples were centrifuged for 15 min at 13,000 rpm. After transfer of the aqueous phase into a new tube, the supernatants were phenol-chloroform extracted once. The RNA was precipitated with two volumes of 100% ethanol and 0.1 volume of 3 M sodium acetate (pH 5.3) for 30 min at −20 °C. After centrifugation, the RNA was resuspended in 100 μl of DEPC-treated water. For each sample, 10 μg of RNA was size-fractionated on a MOPS/formaldehyde gel and Northern blotting was performed. The membranes were probed with 32P-labeled gene-specific PCR probe in ULTRAhyb solution (Ambion, Austin, Texas, United States) and quantified using ImageQuant software for 15 min, exposed to a phosphor-imager screen that was read with a Storm imaging system and quantified using ImageQuant software (Amersham Biosciences, Piscataway, New Jersey, United States). The mRNA levels of the Hgq quantitative trait genes were compared against act1 mRNA as a loading control.

Cloning alleles of quantitative trait genes. The Hgq and Htg alleles of MKT1-End3 (pHS07 and pHS08; see Figure S1) and RH02 (pHS09 and pHS10; see Figure S2) were cloned using a modified gap repair [41] technique. To gap repair cloned alleles of each gene, primers corresponding to the TERT primer (pHS16) forward primer were used to amplify from genomic DNA 0.5-kb regions of the open reading frames of the genes upstream and downstream of the Hgq QTR. The PCR products were then purified using a Qiagen PCR Purification kit and introduced into ura3 strains (S19 and YM791, Table S4) along with SmaI-digested pRS316 plasmid. Circularization and 0.1% SDS for 5 min and twice again with 0.1 SSC and 0.1% SDS for 15 min, then a phosphor-imager screen was read with a Storm imaging system and quantified using ImageQuant software.

Homologous coding allele replacements of the RH02 3’UTR. RH02 3’UTR allele replacements were performed by deleting the RH02 3’UTR, spanning 2 bp upstream of the RH02 stop codon to 208 bp upstream of the TOP2 start codon in strain S705, with a FCY1-NatMX4 cassette (in pBN33, see Figure S5), selecting for Nat sensitivity. The RH02 3’UTR::FCY1-NatMX4 gene in strain S705 was confirmed by PCR into the RH02 3’UTR::FCY1-NatMX4 strain were introduced (i) a 1,172-bp BglII-BglII fragment and (ii) a 1,289-bp BglII-BglII fragment of the RH02 3’UTR::FCY1-NatMX4 cassette into strain S3207 by the cloned RH02 3’UTR::FCY1-NatMX4 DNA was selected by replica plating the colonies growing on YPD+G418 plates. The RH02 3’UTR::FCY1-NatMX4 strains were checked for Kan sensitivity and temperature resistance, and were confirmed by PCR and DNA sequencing.

Supporting Information

Figure S1. Restriction Map of Plasmids pHS07/pHS08 Showing the Cloned Fragment and Restriction Enzymes Used to Obtain MKT1 Fragments

Figure S2. Restriction Map of Plasmids pHS09/pHS10 Showing the Cloned Fragment and Restriction Enzymes Used to Obtain RH02 3’UTR Fragments

Figure S3. Map of Plasmid pSA11 with CEN and NatMX4 Cassette

Figure S4. Restriction Map of Plasmid pBN49

Figure S5. Restriction Map of Plasmid pBN35

Figure S6. Map of Plasmid pBN98

Table S1. MKT1, END3, and RH02 RHA Data

Table S2. MKT1 Coding/Noncoding Polymorphism Transfers
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