Meiotic interactors of a mitotic gene *TAO3* revealed by functional analysis

of its rare variant

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Running title:

Role of mitotic gene *TAO3* in meiosis

Keywords:

Rare variant, *TAO3*, meiosis, transcriptome profiling, allelic variant

29   **ABSTRACT**

Studying the molecular consequences of rare genetic variants has the potential of identifying

novel and hereto uncharacterized pathways causally contributing to phenotypic variation.

Here we characterize the functional consequences of a rare coding variant of *TAO3*,

previously reported to significantly contribute to sporulation efficiency variation in

Saccharomyces cerevisiae. During mitosis *TAO3* interacts with *CBK1*, a conserved NDR

kinase and a component of RAM network. The RAM network genes are involved in

regulation cell separation and polarization. We demonstrate that the role of the rare allele

TAO3(4477C) in meiosis is distinct from its role in mitosis by being independent of *ACE2*,

which is a RAM network target gene. By quantitatively measuring cell morphological

dynamics and conditionally expressing *TAO3(4477C)* allele during sporulation, we show that

TAO3 has an early role in meiosis. This early role of *TAO3* coincides with entry of cells into

meiotic division. Time-resolved transcriptome analyses during early sporulation phase

identified regulators of carbon and lipid metabolic pathways as candidate mediators. We

experimentally show that during sporulation the *TAO3* allele genetically interacts with *ERT1*

and *PIP2*, the regulators of tricarboxylic acid cycle and gluconeogenic enzymes, respectively.

We thus uncover meiotic functions of *TAO3*, a mitotic gene and propose *ERT1* and *PIP2* as

novel regulators of sporulation efficiency. Our results demonstrate that study of causal effects

of genetic variation on the underlying molecular network has the potential to provide more

extensive comprehension of the pathways driving a complex trait. This can help identify

prospective personalized targets for intervention in complex diseases.

INTRODUCTION

The 'common disease, common variant' rationale of genome-wide association studies is being challenged owing to limited fraction of disease heritability explained by the mapped common variants (Manolio *et al.* 2009; Zuk *et al.* 2014). One of the potential contributors to this 'missing' heritability has been suggested to be the potential effects of rare variants which are not considered (Saint Pierre and Génin 2014). This view has been substantiated by the identification of rare variants carrying a considerable risk for autism, schizophrenia and epilepsy (Stankiewicz and Lupski 2010). Thus characterizing the functional role of rare variants associated with complex diseases has the potential for revealing new biology and providing opportunities for treatment (Cirulli and Goldstein 2010; Zuk *et al.* 2014). Even though multiple variants for various diseases have been mapped, they have not been able to provide targets for treatment. This is because firstly, many variants have been mapped in regulatory or non-coding regions, therefore the affected gene is not known. Secondly, even when a variant is in a coding sequence but if the affected gene is not well characterized for the phenotype, the variant may be of little use. This makes characterization of either common or rare genetic variants laborious. Hence the causal path connecting a variant to the phenotype is usually unknown. Thus to understand the causal path of a variant, it is important to identify the mediating molecular pathways. Identifying these mediating pathways has the potential to greatly expand the set of possible targets for molecular intervention (Gagneur *et al.* 2013).

Yeast sporulation efficiency is a complex trait and many causative polymorphisms have been mapped in sporulation genes such as *IME1*, an initiator of meiosis (Gerke *et al.* 2009) and *RIM15*, a glucose-sensing regulator of meiosis (Lorenz and Cohen 2014). However a polymorphism each was identified in two genes with functional annotations described for mitotic growth only. These polymorphisms were in *MKT1*, a putative RNA-binding protein and *TAO3*, a putative scaffolding protein (Deutschbauer and Davis 2005). These novel non-synonymous polymorphisms, *MKT1(89G)* and *TAO3(4477C)* were identified in a high efficiency sporulating SK1 strain while the low efficiency S288c strain had *MKT1(89A)* and *TAO3(4477G)* (Deutschbauer and Davis 2005). In our previous work we determined that *MKT1(89G)* variant increased the sporulation efficiency by genetically interacting with regulators of mitochondrial retrograde signaling and nitrogen starvation during sporulation (Gupta *et al.* 2015). Tao3 encodes a highly conserved scaffolding protein that is a component

of the RAM (Regulation of Ace2p activity and cellular Morphogenesis) signaling network.

Tao3 is required for activation and localization of an NDR protein kinase Cbk1, another

essential component of RAM network (Du and Novick 2002; Hergovich *et al.* 2006). This

RAM signaling network including Cbk1, Hym1, Kic1, Mob2 and Tao3 contributes to various

important processes of mitotic cellular growth. Regulation of transcription factor Ace2 by the

RAM network is critical for cell separation and polarized growth (Nelson *et al.* 2003). Ace2

peaks early in mitosis and is involved in G_1/S transition (Spellman *et al.* 1998). This RAM

network also known to regulate cellular progression through a Ace2-independent pathway

(Bogomolnaya *et al.* 2006). However none of these *TAO3* mitotic interactions provide clues

for its role in the developmental process of meiosis and sporulation.

Here we characterized the functional role of *TAO3(4477C)* in sporulation efficiency variation

by elucidating the molecular pathways linking this mitotic gene in meiosis. We compared

phenotypes of a pair of S288c-background strains differing only for the casual *TAO3*

polymorphism. Genome-wide transcriptional dynamics during sporulation of these strains

identified candidate mediator genes. Allele-specific genetic interaction assay between these

candidate genes and casual *TAO3* allele identified the regulators of tricarboxylic acid cycle

and gluconeogenic enzymes as casual and novel regulators of sporulation efficiency.

103   **RESULTS**

Role of causative allele of *TAO3* in sporulation efficiency variation

Analysis of *TAO3* sequence of 38 *S. cerevisiae* strains in the Saccharomyces Genome

Resequencing Project (SGRP) database (Liti *et al.* 2009) and 24 strains in Saccharomyces

Genome Database (SGD), showed that *TAO3(4477C)* allele of SK1 strain was a rare variant

(minor allele frequency = 1.6%, Figure 1A). Deutschbauer and Davis (2005) mapped this rare

variant as a causal allele for increased sporulation efficiency in a cross between S288c (low

sporulating) and SK1 (high sporulating) strains and introduced it in the S288c background to

construct allele replacement strain YAD331. This allele replacement strain differed from the

parental S288c strain (*TAO3(4477G)*) only for this variant. A clean allele replacement strain,

termed as "T strain" was constructed from YAD331 (see Methods) and its sporulation

efficiency at 48h was reconfirmed to be three-fold higher than the S288c strain (termed as "S

strain", $P = 1.8 \times 10^{-10}$, pair test in Methods, Figure 1B). This fold-difference remained

constant even after a week of incubation (Figure 1C). Studying the progression of meiotic

phases showed that the T strain initiated meiosis within 12h (Figure 1D-E). Quantitative

comparison of the time to initiate meiosis and the rate of transition from G_1/G_0 into Meiosis I

stage showed significant difference between the T and S strains (Figure 1D-E, Figure S1).

This suggested that *TAO3(4477C)* affected entry of the T strain cells initiating meiosis within

12h in sporulation. To resolve when during this 12h time phase *TAO3(4477C)* affects the

phenotype, this endogenous allele in the T strain was placed under a tetracycline-responsive

promoter (P_{Tet}-*TAO3(4477C)* strain, see Methods). In the absence of tetracycline analogue,

doxycycline, P_{Tet}-*TAO3(4477C)* strain showed higher expression of *TAO3(4477C)* relative to

its expression in the S strain (Methods, Figure S2). Addition of 2μg/ml doxycycline

significantly reduced *TAO3* expression level, making it equivalent to the S strain (Methods,

Figure S2). Concomitantly in the absence of doxycycline, the P_{Tet}-*TAO3(4477C)* strain

showed high sporulation efficiency (Figure 1F). In presence of doxycycline for the entire 48h

in sporulation medium, the sporulation efficiency of the P_{Tet}-*TAO3(4477C)* strain was

equivalent to the S strain (Figure 1F). This suggested that high *TAO3(4477C)* expression was

required for the high sporulation efficiency phenotype. We next reduced *TAO3(4477C)*

expression for specific shorter time-periods in the sporulation medium. Sporulation efficiency

of the P_{Tet}-*TAO3(4477C)* strain was equivalent whether doxycycline was present only for the

first 6h or for 48h and this efficiency was equivalent to the S strain (Figure 1F). However

when doxycycline was present only for the first 3h, the P_{Tet}-*TAO3(4477C)* strain showed a

slight but significant difference ($P = 0.02$) in sporulation efficiency compared to S strain

(Figure 1F). This showed that *TAO3(4477C)* allele affected sporulation efficiency within the

first 6h in sporulation.

Role of *TAO3* in meiosis is distinct from its role during mitosis

Varying the gene expression of *TAO3(4477C)* affected the sporulation efficiency phenotype.

Hence to identify the molecular pathways affected by this causative allele, we studied the

global gene expression dynamics during sporulation in the allele replacement strains. Time-

resolved transcriptomics of the T and S strains were compared from 0h to 8h30m in

sporulation medium (see Methods). At the initial time point (t = 0h) only 190 out of 6,960

transcripts (~3%) showed differential expression, with an enrichment for a single gene

ontology term iron ion homeostasis ($P = 0.04$, post Holm-Bonferroni corrected, Figure S3).

In contrast 1,122 transcripts (including non-coding SUTs, Table S1) showed statistically

significant differences in gene expression dynamics as a function of time between the two

strains (FDR cut-off 10%, when controlling for expression at t = 0h). While *TAO3* was

amongst the transcripts showing differential expression dynamics during sporulation ($P =$

0.004), none of its mitotic interactors showed differential expression (Figure 2B). However a

few *ACE2*-regulated genes did show differential expression (11 genes labeled green in Figure

2B), so we studied the effect of *ace2Δ* in the T strain and high sporulating SK1 strain. *ACE2*

is known to regulate the budding phenotype (Voth *et al.* 2005) thus both the T and SK1

strains with *ace2Δ* showed clumping. However *ace2Δ* did not affect sporulation efficiency of

either the T or SK1 strain (Figure 2C). Ace2-independent effect of RAM network on cellular

polarization have been observed previously (Nelson *et al.* 2003) therefore it is possible that

this network could still be involved in meiosis.

To determine whether the mitotic interactors of *TAO3* were distinct from its meiotic

interactors, we again used P_{Tet}-*TAO3(4477C)* strain and reduced *TAO3* expression only

during the mitotic growth phase, i.e. in glucose rich (YPD) medium. We observed no growth

difference between P_{Tet}-*TAO3(4477C)* strain with or without doxycycline and the T strain

(Figure S2). In addition there was no effect on sporulation efficiency among the strains

(Figure 1F). These results implied that probably *TAO3(4477C)* allele had a distinct meiotic

role from its established function in mitosis-related processes.

Temporal gene expression profiling predicts *TAO3(4477C)*-specific interactors during

sporulation

We showed that *TAO3(4477C)* had a distinct role in the sporulation processes within first 6h

of sporulation (Figure 1F). Hence we identified by clustering (see Methods), sets of

differentially expressed genes showing early and increasing trend in their expression profiles

in the T strain only. Various sporulation genes including crucial regulators of meiosis,

namely *IME1*, *IME2*, *DMC1* and *NDT80* were enriched ($P = 5.5 \times 10^{-12}$) in a cluster showing

increasing expression (Cluster II) during sporulation in the T strain (Figure 3B, see Methods).

Approximately 50% of Cluster II genes of the T strain showed a similar increasing trend in

the S strain, including *IME1*, *IME2*, *DMC1*, *ECM11* and *NDT80* (Figure S4, Table S2).

Interestingly very few early expressing genes (Cluster I) of the T strain overlapped with the S

(7%, Figure S4). These genes belonged to biological processes that regulated entry into

sporulation, such as carbohydrate metabolic process, ion transport, mitochondrial

organization and cellular respiration (Table 1). Furthermore genes involved in biological

processes like carbohydrate metabolic process and mitochondrial organization showed

repression in the S strain (Table 1, Figure S5). Therefore to study the early effects of the

causal *TAO3* allele, we identified regulators of only those differentially expressed genes that

showed early and increasing expression uniquely in the T strain (Tables S3 and S4).

These regulators were enriched in nutrient metabolism and chromatin modification –

biological processes important for initiation of meiosis (Neiman 2011). A core sporulation

gene *UME6*, which together with *IME1* induces expression of early meiotic genes (Kassir *et*

al. 2003) and is known to regulate other important processes for initiating meiosis (Table 2,

Lardenois *et al.* 2015) was also identified. Interestingly among the regulators of the genes

showing early and increasing expression uniquely in the T strain, we identified 25 upstream

regulators of *UME6* (Figure 4A, Tables S3-5). Among these regulators were *ERT1*, *OAF1-*

PIP2 and *DAL81*. *ERT1*, a regulator of carbon source utilization (Turcotte *et al.* 2010) is

involved in the switch from fermentation to respiration in glucose-limiting conditions (Gasmi

et al. 2014). *OAF1-PIP2* is a protein complex regulating lipid metabolism (Karpichev and

Small 1998). *DAL81* is a regulator of nitrogen degradation pathway (Marzluf 1997).

Interestingly like *UME6*, *OAF1* target genes were repressed in the S strain (Cluster IV, Table

S6). Earlier work in S288c and SK1 strains has shown upregulation of *ERT1*, *PIP2* and

DAL81 in SK1 strain during sporulation (Primig *et al.* 2000). However their deletion in

S288c strain had no effect on its sporulation efficiency (Deutschbauer *et al.* 2002). A few

other interesting candidate genes that were not upstream *UME6* were also identified (Tables

S3 and S4). These included *GAT1*, a regulator of nitrogen metabolism (Ljungdahl and

Daignan-Fornier 2012) and *GAT3*, a regulator of spore wall assembly (Lin *et al.* 2013). We

next tested if the metabolic regulators identified through this analysis were *TAO3(4477C)*-

specific mediating genes during sporulation.

Allele-specific functional validation identifies *TAO3(4477C)*-specific genetic interactors

during sporulation

The candidate genes predicted in the above analysis could be either causal mediating genes

interacting with *TAO3(4477C)* during sporulation or non-mediating consequential genes

associated with only the genotype or the phenotype. To identify only the causal mediating

genes, we used a genetic model described previously (Figure 4B, Gupta *et al.* 2015).

According to this model if a gene is associated only with the genotype and not with the

phenotype or is expressed as a consequence of the phenotype, its deletion would not affect

the T strain phenotype. If a gene had an independent role in sporulation phenotype, its

deletion will result in both a reduction in phenotype and an additive effect, irrespective of the

genetic background. Any significant deviation from this expectation would imply dependence
on the genotype with epistasis being an extreme case. In this scenario deleting the gene in T
strain would affect the phenotype while deleting the same in the S would not have an effect
on the phenotype, making it a causal mediating gene. While *gat1Δ* had no effect on
sporulation efficiency of the T strain, *ert1Δ*, *pip2Δ* and *gat3Δ* significantly reduced the mean
sporulation efficiency in the T strain by about 1.5-fold ($P = 2.1 \times 10^{-12}$, $P = 6.1 \times 10^{-13}$, $P =$
9.6×10^{-10} respectively, pair test in Methods, Figure 4C). Significant interaction terms were
obtained between the genetic backgrounds (S and T) and *ert1Δ* and *pip2Δ* ($P = 2.3 \times 10^{-4}$, $P =$
0.04, see Methods) but not for *gat3Δ*. This showed that the effect of *ert1Δ* and *pip2Δ* on
sporulation efficiency was specific to *TAO3(4477C)*, making them causal mediating genes.
GAT1 and *GAT3* were non-mediating genes, the former associated with the genotype only or
a sporulation-consequential gene and the latter affected sporulation independent of the
genotype. Therefore genetic and functional validation using this model identified true causal
genes, namely *ERT1* and *PIP2*, mediating the effect of the allelic variant of *TAO3* on
sporulation efficiency.
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235   **DISCUSSION**
Strong effects on phenotypic variation have been observed as a consequence of rare coding
variants (Cohen *et al.* 2004; Cohen *et al.* 2005). Tao3 is conserved from yeast to humans
(Hergovich *et al.* 2006) and had been functionally annotated solely for mitotic cell division
(Du and Novick 2002; Nelson *et al.* 2003) until it was mapped for sporulation efficiency
variation (Deutschbauer and Davis 2005). In this study we identify *ERT1* and *PIP2* as the
TAO3(4477C)-dependent mediators contributing to efficient meiosis. These genetic
interactors of *TAO3(4477C)* are distinct from the mitotic interactors of *TAO3(4477G)*. In this
study we identify their novel regulatory role in sporulation efficiency.
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Acetate is the sole non-fermentable carbon source available to yeast during sporulation in
laboratory conditions. During sporulation, this acetate gets internalized into the tricarboxylic
acid (TCA) and glyoxylate cycle. Gluconeogenesis utilizes the TCA cycle intermediates and
synthesizes storage carbohydrates like trehalose that is utilized during late sporulation
processes (Ray and Ye 2013). Hence TCA, glyoxylate and gluconeogenic metabolic
processes are crucial for sporulation to proceed since reduced flux through these pathways
decreases sporulation efficiency (Aon *et al.* 1996). Moreover the genes encoding the crucial

enzymes of these metabolic processes such as *PFK1*, *CIT1* and *CIT2* are essential for
sporulation (Deutschbauer *et al.* 2002). *ERT1* and *PIP2* are known to regulate these
metabolic enzymes (Baumgartner *et al.* 1999; Gasmi *et al.* 2014). Taken all together, our
results suggest that the rare sporulation-associated *TAO3(4477C)* allele by interacting with
regulators of TCA cycle and gluconeogenic enzymes, modulates the metabolic flux early
during sporulation to result in a better sporulation efficiency.

IME1 acts as a bottleneck for sporulation decision pathway. Lorenz and Cohen (2014)
observed that many sporulation-associated natural polymorphisms have been identified in
genes upstream or interacting with this input/output gene *IME1*. One of such polymorphic
genes *RIM15* is a nutrient-sensing regulator of *IME2*. While *TAO3* and *MKT1* (Gupta *et al.*
2015) do not directly regulate *IME1*, we show that variants in these two genes regulate early
upstream metabolic processes that impinge on *IME1*. Thus our study provides support for the
hypothesis that genes surrounding the signal transduction bottlenecks are reservoirs for
accumulating causal genetic variants.

During mitosis Tao3 localizes to polarized bud sites (Nelson *et al.* 2003). Further
determination of co-localization of *TAO3(4477C)* with membrane-associated *ERT1* and beta-
oxidation regulators *OAF1-PIP2* will give interesting clues of its function during sporulation.
Similar to other scaffolding proteins like *Fry* (*Drosophila*) and *SAX-2* (*C. elegans*), Tao3 has
multiple conserved Armadillo-like repeats (Hergovich *et al.* 2006) and the causal sporulation
variant is present in one of them. Tao3(1493E) physically interacts with the RAM network
proteins in rich growth conditions. It would be interesting to determine binding partners of
Tao3(1493Q) during sporulation and if the variant affects the binding of this putative
scaffolding protein. Additionally a few genes enriched for iron metabolism were
differentially expressed during growth phase prior to incubation in the sporulation medium (t
= 0h). It would be interesting to study whether this metabolic effect of *TAO3* also plays a role
in sporulation.

Even if the basic cellular network of an organism is known, it is crucial to understand how
natural genetic variation and stress conditions modulate the molecular interactions within this
network resulting in differences in phenotypic outcomes (Gasch *et al.* 2016). Studies, such as
this work, aimed at understanding the molecular consequences of genetic variation are
especially important in the field of personalized medicine to make more reliable predictions

regarding the functional consequences of an individual's genotype on disease predisposition
and treatment (Burga and Lehner 2013).
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289   **MATERIALS AND METHODS**
Yeast strains and media
The yeast strains were grown in standard conditions at 30˚C in YPD (1% Yeast extract, 2%
Bacto peptone, 2% dextrose). Allele replacement strain YAD331 (Deutschbauer and Davis
2005) was a S288c-background diploid strain containing the homozygous causative
sporulation polymorphism *TAO3(4477C)*. Whole-genome resequencing of YAD331 with
S288c strain as the reference strain identified two additional polymorphisms (Figure S6,
Table S7). Three consecutive backcrosses were performed between the haploid derivative of
YAD331 and the haploid reference strain (S288c) to remove these secondary polymorphisms.
After the backcrosses, the sole genetic difference between the reference S288c strain and the
backcrossed allele replacement strain was at *TAO3(G4477C)* position, which was confirmed
by performing PCR-based sequencing 650bp up and downstream around the two secondary
polymorphisms and the *TAO3* polymorphic nucleotide. This backcrossed strain was
diplodized to make it homozygous at *TAO3(4477C)* position and was termed as "T strain" in
this study. The diploid parental strain S288c was termed as "S strain" in the study. All gene
deletions in the study were made in the haploids of T and S strains except the ones made in
SK1 strain (Table S8). Deletions were performed and verified as described previously
(Goldstein and McCusker 1999; Gietz and Woods 2002). The haploid strains were diplodized
using pHS2 plasmid (containing a functional *HO*) and mating type were confirmed by
performing MAT PCR (Huxley *et al*. 1990). All the experiments in this study were
performed using the diplodized parent strains and their diploid derivatives. For replacing the
endogenous *TAO3* promoter (-150 to -1bp upstream start site) in the T strain with a
tetracycline-responsive promoter, a *tetO$_7$*-based promoter substitution cassette containing
kanMX4 was amplified from the plasmid pCM225 (Bellí *et al*. 1998b). The diploid T strain
with this *tetO$_7$*-based cassette is termed P$_{Tet}$-*TAO3(4477C)* strain. The primers for
sequencing, deletions and their confirmations are listed in Table S9.
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Phenotyping
Sporulation efficiency estimation at 48h, progression through meiotic landmark events
Meiosis I (MI) and Meiosis II (MII) and its quantitation was done as described previously

(Gupta *et al.* 2015). For quantitation of meiotic landmarks in the T strain, parametric curves

assuming delayed and 1st order kinetics were fitted to the DAPI-stained meiotic progression

time course data and fitting uncertainties were estimated by bootstrapping (File S1). Cell

cycle progression data for S288c and SK1 strains was taken from Gupta *et al.* (2015) (Figure

1D-E). Conditional expression of *TAO3(4477C)* was performed by constructing P_{Tet}-*TAO3*

strain (details in File S1), which was responsive to tetracycline-analogue doxycycline (Bellí

et al. 1998a; Bellí *et al.* 1998b). Doxycycline (2μg/ml) was added in growth and sporulation

media to decrease the expression of *TAO3* gene. For each strain, a minimum of three

biological replicates was used and the experiment was carried out a minimum of two times.

Approximately 300 cells were counted per replicate. Fold difference was calculated as the

ratio of mean sporulation efficiencies of the two strains A and B when the sporulation

efficiency of A is greater than of B. Growth curve analysis was performed for individual

strains grown in YPD in 96-well plates. Cells were grown overnight in YPD to saturation, re-

inoculated in YPD in transparent 96-well plates with a starting OD_{600} of 0.01 and grown with

shaking at 30˚C for 24h in Tecan Infinite M200 microplate reader. Doubling times were

calculated from OD measurements of liquid cultures at a wavelength of 600 nm in the Tecan

reader. For each strain, four technical replicates for each of the three biological replicates

were used. Raw sporulation efficiency values are given in Table S10.

Statistical test for calculating sporulation efficiency

For comparing sporulation efficiency, two statistical tests were used: the pair test and the

interaction test. The pair test tests the null hypothesis that the two given strains (S and T)

have the same sporulation efficiency.

The number $y_{i,k}$ of sporulated cells (4-nuclei count) among the total number of cells $n_{i,k}$ of

strain *i* in replicate experiment *k* was modeled with a quasi-binomial generalized linear model

using the *logit* link function and subject to a common log-odd ratio β_i between replicates, i.e.:

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$$\log\left(\frac{\mu_{i,k}}{n_{i,k}-\mu_{i,k}}\right) = \beta_i \text{ for all } k,$$

where $\mu_{i,k} = E\left(y_{i,k}\right)$.

The pair test tests the null hypothesis of equality of log odd-ratios for two strains i and j, i.e.

$H_0 : \beta_i = \beta_j$.

In the case of the S and T strains, the interaction test tests the null hypothesis that the effect

of mutation A is independent of the effect of mutation B taking the T strain as reference

background. This test thus compares four strains: mutation A only, mutation B only, both A

and B and neither A nor B (T strain). Here the S strain was considered as a T strain mutated

for *TAO3(4477)*. For every interaction test, we considered the dataset of the four strains of

interest and fitted a quasi-binomial generalized linear model using the *logit* link function and

subject to:

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$$\log\left(\frac{\mu_{i,k}}{n_{i,k} - \mu_{i,k}}\right) = \beta_0 + \beta_A A_i + \beta_B B_i + \beta_{A,B} A_i B_i \text{ for all } k,$$

where A_i and B_i are indicator variables of the mutations A and B in strain i respectively. The

interaction test tested the null hypothesis that the odd ratio of sporulation in the double

mutant equals the product of the odd ratios of each mutation, i.e. $H_0 : \beta_{A,B} = 0$.

Both the pair test and the interaction test were implemented in the statistical language R with

the function *glm()* assuming a constant variance function fitted by maximizing the quasi-

likelihood and using the t-test on the tested parameters (Gupta *et al.* 2015).

Whole genome gene-expression profiling

Sporulating yeast cell collection at 0h, 30m, 45m, 1h10m, 1h40m, 2h30m, 3h50m, 5h40m

and 8h30m (logarithmic time-series), RNA isolation and cDNA preparation were performed

as previously described (Xu *et al.* 2009). Samples were hybridized to *S. cerevisiae* yeast

tiling array (Affymetrix, Cat# 520055). Arrays at each time point for both the strains were

normalized together using *vsn* normalization method (Huber *et al.* 2002). For qPCR, aliquots

of cDNA were used in real-time PCR analyses with reagents from Kapa SYBR fast Universal

qPCR master mix (Kapa Biosystems) in the Eppendorf Real-time PCR system according to

manufacturer's protocol. For each strain, four technical replicates for each of the three

biological replicates were used. The primers used are given in Table S9.

Whole genome gene-expression analysis

Within each strain, the $\log_2$ expression values obtained were smoothed using *locfit* at optimized bandwidth parameter $h = 1.2$ (Figure S7), base transformed for each transcript by subtracting the expression value at each time point from the baseline value at time point t = 0h (t_0, Table S11). This $\log_2$ fold change value with respect to t_0 is described as "expression" throughout the manuscript. For identifying the genes showing temporal differential expression between the T and S strains (Table S1), method implemented in EDGE software was used, which calculated statistically significant changes in expression between the T and S strains over time (Storey *et al.* 2005). The differentially expressed genes were clustered according to their temporal expression patterns using time abstraction clustering algorithm implemented in the TimeClust software (Magni *et al.* 2008, see File S1). Four major clusters were identified in each strain: Cluster I (early trend), Cluster II (increasing trend), Cluster III (late trend), Cluster IV (repressing trend) (Table S2). The transcription factors regulating a cluster of genes were extracted using the YEASTRACT database (Teixeira *et al.* 2013). Only those transcription factors were considered as candidate genes whose target genes were significantly enriched in the corresponding cluster ($P \leq 0.05$, odds ratio ≥ 1.5). YEASTRACT database was also used to obtain the regulation matrix of yeast for identifying target genes of regulators in this study such as *UME6*. Target genes for *ACE2* were obtained from Nelson *et al.* (2003). Significantly enriched Gene Ontology terms by biological process (Bonferroni corrected $P < 0.05$, Table 1) were obtained from SGD Yeastmine (Balakrishnan *et al.* 2012).

Data availability

The array data for the T strain has been deposited in ArrayExpress (http://www.ebi.ac.uk/arrayexpress/) with accession number E-MTAB-3889. The whole genome sequence data for the T strain has been deposited in the European Nucleotide Archive (http://www.ebi.ac.uk/ena/) with the accession number PRJEB8698. The rest of the data are available as Supporting Information. The array data and the whole genome sequence data for the S strain were downloaded from Gupta *et al.* (2015). *TAO3* gene sequence data for SGRP strains (Liti *et al.* 2009) was downloaded from (http://www.moseslab.csb.utoronto.ca/sgrp/). Additional 24 *TAO3* sequences were downloaded from Saccharomyces Genome Database (SGD, http://www.yeastgenome.org/cgi-bin/FUNGI/alignment.pl?locus=YIL129C, date accessed: 01 March 2016).

ACKNOWLEDGEMENTS

We thank Manu Tekkedil for help with whole genome sequencing sample preparation and Allan Jones and Gyan Bhanot for helpful comments. This research was supported by Tata Institute of Fundamental Research intramural funds and Department of Biotechnology grant BT/PR14842/BRB/10/881/2010 (H.S.); Bavarian Research Center for Molecular Biosystems and Bundesministerium für Bildung und Forschung through the Juniorverbund in der Systemmedizin "mitOmics" grant FKZ 01ZX1405A (J.G.); the National Institutes of Health, Deutsche Forschungsgemeinschaft and a European Research Council Advanced Investigator Grant (L.M.S.). The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

426   **TABLES**

427   **Table 1. Functional GO categories of clusters in the T and S strains**

Cluster	Functional GO category	Genes
Early in T strain (Cluster I)	Carbohydrate metabolic process	*DOG1, YPI1*
	Ion transport	*AVT4, DAL5*
	Mitochondrion organization	*PPE1, UPS3*
	Cellular respiration	*COX5B*
Early in T strain (Cluster I) repressed in S strain (Cluster IV)	Carbohydrate metabolic process	*ALG6, DEP1, DOG1, TPS3, YPI1*
	Mitochondrial organization	*ATG33, COX20, PPE1, UPS3*

429   Comparison of functional GO categories of differentially expressed genes in the T strain

430   clusters with the S strain. See Table S2 for the full list of genes in each cluster.

432   **Table 2. Functional GO categories of regulators of clusters in the T and S strains**

Functional GO category	Regulators	*P* value
Carbon metabolism	*ERT1, OAF1, PIP2, MIG1, MIG2*	1.9×10^{-6}
Nitrogen catabolite regulation	*DAL81, DAL82, GAT1, UME6*	1.7×10^{-5}
Chromatin modification	*ISW1, PHO2, PHO4, UME6, OAF1, XBP1, SIF2, RSC2*	1.4×10^{-5}

434   Functional GO classification of the regulators of the differentially expressed genes showing

435   early and increasing expression only in the T strain. See Tables S3 and S4 for the full list of

436   genes.

                                    FIGURE LEGENDS

Figure 1. Role of *TAO3* in sporulation efficiency

(A) Comparison of genomic sequence of *TAO3* (4,441-4,500) across the SGRP collection

(Liti *et al.* 2009). The 4,477th position of *TAO3* consists of the sporulation causative variant

where identical nucleotides are indicated by the same color. Identity indicates the percentage

match between the nucleotides in the shown region of the gene. The strains are ordered

according to their mean sporulation efficiency (Tomar *et al.* 2013): high (60-100%),

intermediate (10-60%), low (0-10%) and ND (not determined).

(B) Bar plots represents the mean sporulation efficiency after 48h of the SK1, T and S strains.

The sporulation efficiency data is indicated as circles.

(C) Line graphs represent the mean sporulation efficiency of the S, T and SK1 strains

measured till saturation, i.e. till sporulation efficiency did not vary for 3 consecutive days.

(D) Percentage of 1-, 2- and 4-nuclei states of the T strain (y-axis) versus time in sporulation

medium (x-axis). 1-nucleus stage is indicated as red circles (G_0/G_1 phase), 2-nuclei state as

yellow circles (completion of Meiosis I, MI phase) and blue circles is 4-nuclei stage

(completion of Meiosis II, MII phase).

(E) Bootstrap distribution of the time to initiate meiosis and the rate of transition from G_1/G_0

into MI, estimated from time courses in (D). See Methods for details.

(F) Conditional expression of *TAO3(4477C)* during sporulation in P_{Tet}-*TAO3(4477C)* strain

(indicated as P_{Tet}). Y-axis is the mean sporulation efficiency in 48h. No doxycycline in

growth (YPD) or spo (YPA + sporulation) medium is depicted as "-" condition on x-axis and

addition of doxycycline is depicted as "+" in that condition. "+3h" condition in Spo implies

doxycycline was throughout in the growth medium and in the sporulation medium till 3h

after which cells were sporulated in the absence of doxycycline. "+6h" condition implies

doxycycline was throughout in the growth medium and in the sporulation medium till 6h

after which cells were sporulated in the absence of doxycycline. *P* value was calculated by an

unpaired t-test. Error bars are standard error of mean.

Figure 2. Role of *TAO3* in meiosis is distinct from its role during mitosis

(A) Heatmap showing gene expression of RAM network genes and Ace2-regulated genes in

the T and S strains. Gene names in green show differential expression (data in Tables S1 and

S11).

(B) Bar plots represent the mean sporulation efficiency after 48h of the SK1 and T wild type

(wt) and *ace2Δ* deletion strains. Pair and interaction tests (described in Methods) were

performed to test significance.

(C) Expression profile (log$_2$ fold change t$_0$) of *TAO3* is given in the y-axis for the T (purple)

and S strains (red) and the x-axis denotes the time in sporulation medium (data in Tables S1

and S11).

Figure 3. Global gene expression variation in presence of causative *TAO3* allele

(A) Temporal heat map of meiotic genes in the T and S strains. The gene names shown in

green are differentially expressed in the presence of *TAO3(4477C)*.

(B) The expression profile (log$_2$ fold change t$_0$) for the meiotic landmark genes is given in the

y-axis and the x-axis denotes the time in sporulation medium. Red line represents the

expression profile of the respective gene in the S strain and blue line is the same in the T

strain.

(C) Heat map of the T and S strains showing differentially expressed gene across time within

each cluster. Each row represents a single gene and columns are time points of each strain

(for gene list in each cluster see Table S2). The order of genes in the two strains is based on

the clustering of the T strain. Functional GO categories of genes in each cluster are shown on

left. The boxplots shown on right represent the average expression profile of each cluster in

the T and S strains. The number of genes in each cluster in a strain is indicated in brackets.

Figure 4. Identifying candidate genes mediating the allele specific effects of *TAO3*

during sporulation using the temporal gene expression data

(A) Regulatory network of candidate genes predicted to mediate the effects of *TAO3(4477C)*

in sporulation. The candidate mediating genes are shown as bigger nodes (large circles) with

their target genes (small circles) connected to them as straight lines. The box contains the

protein network interactions of the candidate genes with the core sporulation gene *UME6*,

obtained from YEASTRACT (see Methods). Colors inside the nodes were calculated as an

average of the first six time points in sporulation (early phase). For complete list of

interacting genes and their expression values, see Tables S5 and S11, respectively.

(B) Genetic model for functional validation of allele-specific interactors mediating

sporulation efficiency variation. Wild type effect comparison of the two alleles A1 and A2 of

YFG1 gene is shown inside the box. A1 is associated with high sporulation efficiency (wild

type genotype and phenotype shown) and A2 is associated with low sporulation efficiency

(wild type genotype and phenotype shown). Genetic interaction of these *YFG1* alleles with

candidate mediating genes (*YFG2*) is represented – (i) Representation of non-mediating gene

associated with genotype only or is a consequence of the phenotype since *yfg2Δ* in the

presence of A1 does not affect the wild type phenotype of A1; (ii) Representation of non-

mediating gene associated with the phenotype independent of the allele since *yfg2Δ* in the

presence of both A1 and A2 lowers (low) the phenotype; (iii) Representation of causal

mediating gene since *yfg2Δ* only in presence of allele A1 lowers the phenotype and in the

presence of allele A2 does not change the wild type phenotype of A2.

(C) Bar plots represent the mean sporulation efficiency after 48h of the T and S wild type

(wt) and *ert1Δ*, *pip2Δ* and *gat3Δ* strains. Pair and interaction tests (see Methods) were

performed to test significance.

515   **SUPPORTING INFORMATION**

File S1. Detailed methods

Table S1. Differentially expressed genes between the T and S strains with their *P* and *Q*

values calculated using EDGE

Table S2. Genes in each cluster using TimeClust

Table S3. Transcription factors regulating unique early (Cluster I) genes of the T strain

Table S4. Transcription factors regulating unique increasing (Cluster II) genes of the T strain

Table S5. Differentially expressed target genes of regulators of candidate genes mediating

the affect of *TAO3*

Table S6. Transcription factors regulating unique repressing (Cluster IV) genes of the S

strain

Table S7. Whole genome resequencing results for the *TAO3* allele replacement strain

Table S8. Strain list

Table S9. Primer list

Table S10. Raw sporulation efficiency values

Table S11. Smoothed expression data, base transformed with respect to t_0 for the T and S

strains

Figure S1. Mathematical modeling to identify stages of meiosis affected by *TAO3* causal

allele

Figure S2. Growth phenotype and *TAO3* expression in P_{Tet}-*TAO3(4477C)* strain

Figure S3. Comparison of global gene expression between the T and S strains at t = 0h

Figure S4. Comparison of genes showing early (Cluster I) and increasing trend (Cluster II)

between the T and S strains

Figure S5. Genes having early expression in the T strain show expression at later time points

or repressed in the S strain

Figure S6. Whole genome resequencing of *TAO3* allele replacement strain (YAD331,

(Deutschbauer and Davis 2005) in comparison to the S288c reference strain

Figure S7. Smoothing of normalized temporal data using *locfit*

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