Quantitative trait locus mapping of osmotic stress response in the wheat pathogen

Zymoseptoria tritici

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Abstract

Osmotic stress is a ubiquitous and potent stress for all living organisms. Fungal molecular

responses to salt stress in well studied model yeast species highlighted the importance of

the conserved HOG MAP kinase pathway. However, fewer studies have investigated the

genetic basis of salt tolerance in filamentous fungi. The main aim of this study was to

identify regions of the genome associated with tolerance to potassium chloride in the wheat

pathogen *Zymoseptoria tritici.* A secondary aim was to identify candidate genes affecting

salt tolerance within the most promising chromosomal regions. We achieved these aims

with a QTL mapping study using offspring from two crosses grown in vitro in the presence or

absence of osmotic stress imposed by 0.75M KCl. We identified significant QTLs for most of

the traits in both crosses. Several QTLs overlapped with QTLs identified in earlier studies for

other traits and some QTLs explained trait variation in both the control and salt stress

environments. A significant QTL on chromosome 3 explained variation in colony radius at 8

days post inoculation (dpi) in the KCl environment as well as KCl-tolerance affecting colony

radius at 8 dpi. The QTL peak had a high LOD and encompassed an interval containing only

36 genes. Six of these genes present promising candidates for functional analyses.

Keywords:
Salt stress; plant pathogen; fungi; *Mycosphaerella graminicola*, QTL

1. Introduction

Osmotic stress is a common and potent environmental stress that can limit growth and reproduction in many organisms. At the cellular level osmotic stress results in changes to the movement of water into or out of the cell through osmosis. To counteract osmotic stress, organisms need to accumulate compatible solutes (e.g. glycerol, erythritol, arabitol, mannitol and trehalose) – that can balance osmosis and provide turgor pressure (Araújo et al., 2020; Kogej et al., 2007). In fungal pathogens of plants, the pathogen may be exposed to a relatively high salt concentration (0.5M) within the plant (see Araújo et al., 2020) as well as experience changes in salt and osmolyte concentrations as plant tissues degrade (Jacob et al., 2015). The genes and pathways involved in responding to the resulting osmotic stresses may also affect plant-pathogen interactions and virulence. For example, turgor pressure which is mediated by changes in vacuole size and number is important in both plant tissue penetration and response to osmotic stress (Richards et al., 2010). Voltage-gated chloride channels, which are involved in regulation of cell volume, also influence virulence in fungal pathogens of both plants and humans (Cañero and Roncero, 2008; Zhu and Williamson, 2003). Knowledge of the genetic basis of osmotic sensitivity in plant pathogens can thus provide important insights into the genetic basis of environmental stress responses as well as disease development.

The immediate effects of osmotic stress in fungi include stalled growth, changes in the recruitment of water, collapse of the cytoskeleton, altered plasma membranes and a reduction in glycerol transport (Duran et al., 2010; Hohmann, 2002). Longer-term adaptation to osmotic stress involves modifications of metabolism, plasma membranes, cell wall thickness and cell wall melanisation, as well as an accumulation of mycosporines (Hohmann, 2002; Kogej et al., 2007). Osmotic stress is also likely to activate many genes and pathways involved in the 'general stress response'. For example, up to 10% of the genes that were differentially expressed under different individual stressors were shared across these different treatments (Hohmann, 2002).

Although the key salt stress response pathways and genes have been identified and well characterized in model yeasts, studies in filamentous fungi could provide novel insights into the genes involved in responding to osmotic stress. The HOG pathway is an osmoresponsive

system and one of the most well-understood Mitogen-activated protein (MAP) kinase

pathways (Hohmann, 2002). In yeast, two genes are central components of this pathway –

Pbs2 and Hog1p kinase. In plant pathogens, HOG signalling appears more complex than in

yeast, although it is still poorly understood (Jacob et al., 2015). Several filamentous

pathogenic fungi (including our model *Zymoseptoria tritici*) have multiple (up to 17) histidine

kinases, whereas *S. cerevisiae* has only one (Sln1p) (Jacob et al., 2015). Several studies have

identified transcription factors involved in regulating fungal response to salt stress, as well

as other stressors - including temperature and fungicide stress. For example, in *Aspergillus*

oryzae, GATA transcription factors (TFs) were shown to be important in responding to salt

and temperature stress (Jiang et al., 2021). In the causal agent of Fusarium head blight

(*Fusarium graminearum*), the Nutrient and stress factor 1 (*NSf1*) C_2H_2 zinc fingers protein

plays an important role in salt stress response, sexual-asexual reproduction, vegetative

growth, fungicide sensitivity and pathogenicity (Shi et al., 2021), and a Zn(2)-C6 fungal-type

DNA-binding TF (Fss1) is required for sodium and lithium tolerance (Son et al., 2015). Cell

wall melanisation also appears to confer benefits for osmotic stress tolerance. For example,

in halotolerant fungi cell wall melanisation helps to maintain high concentrations of glycerol

and contain the resulting turgor pressure by strengthening and preventing leakage through

the cell wall (Kogej et al., 2007). These examples coming from a relatively small number of

fungal species demonstrate how the molecular mechanisms of salt response can vary across

taxa and illustrate that additional work is needed to understand how fungi adapt to salt

stress.

To improve our understanding of osmotic stress response in fungi we performed a

quantitative trait loci (QTL) mapping study using the wheat pathogen *Z. tritici* growing under

salt stress. QTL studies have been instrumental in finding candidate genes for multiple traits

in this widespread and damaging wheat pathogen, including melanisation (Lendenmann et

al., 2014), fungicide sensitivity (Lendenmann et al., 2015), morphological switching

(Francisco et al., 2022) and oxidative stress (Zhong et al., 2021), and enabled the cloning and

functional validation of genes affecting virulence (Zhong et al., 2017; Meile et al. 2018),

melanisation (Krishnan et al. 2018), and morphological switching (Francisco et al., 2022).

QTL mapping identifies genomic regions that explain variation in a trait and that are likely to

harbour genes responsible for creating variation in the trait. The aim of this study was to

identify large effect loci associated with tolerance to salt stress in *Z. tritici* and identify

plausible candidate genes within the most promising QTL intervals.

2. Methods

2.1. QTL mapping crosses

Parent and progeny isolates used in this study are from two crosses that were described

previously (Lendenmann et al., 2014). The crosses were made by coinfecting wheat leaves

with the parent strains, with the resulting ascospores collected from infected leaves and

grown in vitro. One cross was between ST99CH1A5 and ST99CH1E4 (herein referred to as

1A5 and 1E4 respectively) and the other was between ST99CH3D7 and ST99CH3D1 (herein

referred to as 3D7 and 3D1 respectively).

2.2. Genotyping

SNP genotype data for the progeny were obtained from a RAD sequence dataset that was

previously produced in our lab (first described in (Lendenmann et al., 2014). In brief, the

genome was cut using the restriction enzyme *Pst*I and the libraries were sequenced on an

Illumina HiSeq2000 with paired-end sequencing. Complete genome sequences of the

parental strains (Croll et al., 2013)(NCBI Biosample: SRS383146 (ST99CH3D1), SRS383147

(ST99CH3D7), SRS383142 (ST99CH1A5), and SRS383143 (ST99CH1E4)) were used to SNP

genotype the parents and offspring.

RADseq processing and variant discovery is described in detail at

https://github.com/jessstapley/QTL-mapping-Z.-tritici. In brief, RADseq reads were trimmed

of adaptors and low-quality sequence using trimmomatic (v0.35). The RADseq reads were

mapped to a reference genome using bwa mem (v0.7.17). Reads from progeny of the

3D7x3D1 cross were mapped to the 3D7 reference genome and progeny from the 1A5x1E4

cross were mapped to the 1A5 reference genome. Variant calling was done using the GATK

Germline Short Variant Discovery pipeline, following their Best Practice recommendations

(https://gatk.broadinstitute.org/hc/en-us/sections/360007226651-Best-Practices-

Workflows). After this we applied the following filters: only biallelic SNPs, parents had

alternative alleles, <50% missing genotypes per marker, <50% missing genotypes per

individual, mean read depth >3 and <30, depth quality (QD)>5, mapping quality >40, minor

allele frequency >0.02 and <0.80, and SNPs in regions where the SNP density is >3 in 10bp

were removed.

2.3. Generating linkage maps

Linkage maps were made using R (R Core Team, 2020), R Studio (RStudio Team, 2020) and

the R package qtl (v1.48-1 (Broman et al., 2003)). As the order and base pair positions of the

markers were known from the reference genomes, we only needed to calculate the

recombination distance between markers to create the linkage map. Before map

construction, we removed putative clones. To do this we calculated genetic similarity using

the 'comparegeno' function and removed individuals with >99% shared genotypes (83

progeny were removed from the 1A5x1E4 cross, 74 progeny were removed from the

3D7x3D1 cross). The linkage map was constructed for each chromosome separately using

the 'estimatemap' function with method 'morgan'. Then we performed multiple rounds of

cleaning to remove error-prone markers that caused map inflation. First, we removed

markers that had an estimated recombination distance above the 99.9% quantile. Then

recombination rate was estimated again and we inspected the maps. In regions of the map

where there was evidence of map inflation and spurious double recombination events, we

visualised these regions in an Integrated Genome Viewer (IGV, https://igv.org) to look at the

mapping alignments. Markers were removed from regions with poor alignments and

evidence of mismapping. A summary of the two linkage maps is provided in Supplementary

Material S1 and the complete maps are available at (https://github.com/jessstapley/QTL-

mapping-Z.-tritici). After applying all these filters, the average marker spacing was 1217 bp

(0.19 cM) and 529 bp (0.14 cM) for the 1A5x1E4 and 3D7x3D1 crosses, respectively.

2.4. Phenotyping

We compared colony growth and melanisation in a control environment and in the

presence of potassium chloride (KCl) to simulate a salt stress environment. Potassium

chloride (KCl) has been used extensively in the study of stress response in fungi because it is

environmentally ubiquitous and lacks the toxic effects of Na^+ (Araújo et al., 2020; Hohmann,

2002). Only non-clonal offspring strains were phenotyped (for 1A5x1E4 n= 259 strains; for

3D7x3D1 n=265 strains). The basic protocols for isolate recovery from -80^0C storage, growth

in vitro and the measurements of colony size and colony melanisation were described

previously (Zhong et al., 2021). In brief, Petri dishes containing Difco potato dextrose agar

(PDA) were inoculated with 200μl of a spore solution (concentration 200 spores/ml) and

grown at 18^0C for 8 and 12 days. In the salt stress environment, a sterile potassium chloride

solution was added to the cooled and sterilized PDA to make a final concentration of 0.75M.

Three replicates were grown in both the control and salt (KCl) environments.

Two traits were measured; colony area and colony grey value, using automated image

analysis as described previously (Lendenmann et al., 2014; Zhong et al., 2021). Grey value

was measured on a scale of 0-255, where darker, more melanised colonies have lower

values (0 = black, 255 = white). Mean colony area and grey value were calculated from

measurements of multiple colonies on each plate and then the mean was calculated across

the three replicate plates to obtain a final mean colony area and mean grey value for each

strain in the two environments. Colony area was converted to a radial size measure by

dividing by π and taking the square root of this value. Measurements were taken at 8- and

12-days post inoculation (dpi). For the analysis we used three different types of data; time

point/age: the colony radius and grey value at 8 and 12 dpi; daily rate measurements: the

change in colony radius or grey value between 8-12 dpi (i.e. the difference in colony radius

between 8 and 12 dpi/difference in days); and tolerance measurements: the ratio of a

measurement (colony radius, grey value, growth rate or melanisation rate) between the KCl

environment and the control environment (i.e. radius at 8 dpi in KCl environment/radius at

8 dpi in control environment, or growth rate in KCl environment/growth rate in control

environment). For the tolerance measurements, values >1 indicate the isolate is more

tolerant to KCl (grew faster under salt stress), and values <1 indicate that the isolate is more

sensitive to KCl. In total there were 18 traits (Table 1).

2.5. QTL mapping

QTL mapping was performed using the R (v 3.6.0) package 'qtl2' (v2_0.24) as described in

detail on github (https://github.com/jessstapley/QTL-mapping-Z.-tritici). We scanned the

genome for a single QTL per chromosome with the '*scan1*' function using a linear mixed

effect model, and we controlled for the relatedness of individuals (i.e. we included a

random polygenic effect) by fitting the kinship matrix as a random effect. Models that take

into account the genetic covariance between individuals can reduce the false discovery rate

in QTL scans and outperform models that do not include this information in the model

(Malosetti et al., 2011). The significance threshold for a QTL peak was determined by

permutation tests (n=1000) and we calculated a Bayes Credible Interval (95%) to identify the

interval size around the QTL peak. We used custom R scripts to extract a list of the genes

within the interval from the annotation (gff) files of each reference genome. The putative

encoded function of each gene was determined in previous analyses (for 1A5 see

(Plissonneau et al., 2018), for 3D7 see (Plissonneau et al., 2016)). The annotations were

obtained using Interproscan (https://www.ebi.ac.uk/interpro/) against multiple protein

databases and then screened using multiple methods (i.e. signalP

https://services.healthtech.dtu.dk/service.php?SignalP-5.0, Phobius

https://phobius.sbc.su.se/) to determine if the encoded proteins contained likely signal

peptides or secreted domains. The putative effect of a SNP in the coding sequence was

determined using SNPeff (http://pcingola.github.io/SnpEff/).

2.6. Identifying putative candidate genes in the 3D7x3D1 cross

Genes within a single QTL interval on chromosome 3 were further investigated to identify

the most likely candidate genes responsible for the phenotype. We identified the

orthologous genes in the 3D1 parent using an analysis performed across the genomes of 19

reference strains (Badet et al., 2020). We inspected the alignments in the region of interest

in IGV. We also blasted the 3D1 gene sequences (including ±200bp flanking regions) against

the 3D7 genome to calculate sequence similarity and identify indels.

2.7. RNAseq data analysis

We analysed RNAseq data that was previously created for the four parental strains during

infection of the susceptible wheat cultivar Drifter (Palma-Guerrero et al., 2017) as well as

cultures grown in vitro on two types of liquid media (yeast sucrose broth (YSB): yeast extract

10 g/L and sucrose 10 g/L, pH 6.8; and carbon-depleted minimal medium (MM), pH 5.8

(Francisco et al., 2019)). The raw sequence reads were downloaded from the Short Read

Archive (SRA, https://www.ncbi.nlm.nih.gov/sra, Bioproject: *invitro* SRP152081, *invivo*

SRP077418). RNAseq data processing and analysis is described in detail at github

(https://github.com/jessstapley/QTL-mapping-Z.-tritici). In brief, reads were trimmed and

then mapped to one of two reference genomes (reads from 3D1 and 3D7 were mapped to

the 3D7 reference genome while reads from 1E4 and 1A5 were mapped to the 1A5

reference genome). Then the number of reads mapping to each gene was counted using

R::Rsubread (Liao et al., 2019) and we tested for differential gene expression between

strains, 3D7 versus 3D1 and 1A5 versus 1E4, using R::EdgeR (Robinson et al., 2010).

3. Results

3.1. Effect of KCl on growth and melanisation

A total of 158,480 and 120,059 colonies from the 1A5x1E4 and 3D7x3D1 crosses,

respectively, were phenotyped (see Table S1 for details). We used paired Wilcoxon tests to

compare mean trait values between the control and KCl treatments (Table S1). In the

presence of KCl, colony radius was smaller at 8 and 12 dpi in both crosses, and the growth

rate was reduced in the 3D7x3D1 cross, but not in the 1A5x1E4 cross, suggesting that

isolates from the 3D7x3D1 cross were more sensitive to KCl. This is consistent with the

greater mean KCl-tolerance found in the 1A5x1E4 cross compared to the 3D7x3D1 cross

(KCl-tolerance in growth: 3D1 = 1.04 versus 3D7 = 0.96, KCl-tolerance in radius at 8 dpi 3D1

= 0.42 versus 3D7 = 0.40, KCl-tolerance in radius at 12 dpi 3D1 = 0.70 versus 3D7 = 0.68). In

the KCl environment we observed less melanisation at both time points and a reduced

accumulation of melanin over time (melanisation rate) in both crosses (Table S1).

3.2. Correlations between traits

All pairwise correlations (Pearson, non-parametric) between independent measurements

are provided in Tables S2 and S3. Excluding traits that are statistically or functionally

coupled, which includes: correlations between two time points of the same trait (i.e. radius

at 8 dpi and 12 dpi in the control), correlations between a rate and the corresponding time

point measurements (i.e. grey value at 12 dpi and melanisation rate in the control), or

correlations between rates and tolerance measurements, we identified many correlations

between independent measurements (Table S2 and S3). Notably, growth rate in the control

environment was correlated with growth rate in the KCl environment in both crosses

(correlation (1A5x1E4) = 0.63 and correlation (3D7x3D1) = 0.62) – thus isolates that grow

faster in a benign environment also grow faster in the presence of salt stress, i.e. they have

a higher intrinsic growth rate. Melanisation rates showed a similar trend in 1A5x1E4 -

isolates that had higher melanisation rates in the benign environment also had higher

melanisation rates under KCl stress (Table S3, correlation = 0.19), but this was not observed

in the 3D7x3D1 cross (Table S2, correlation = 0.07).

We were particularly interested in the relationship between KCl tolerance and melanisation,

as melanin can help protect cells from osmotic stress. In both crosses, the mean grey value

in the control environment at 12 dpi was negatively correlated with KCl-tolerance affecting

growth rate (Tables S2 and S3, 1A5x1E4 correlation = -0.26 and 3D7x3D1 correlation = -

0.43), the relationship at 8 dpi was also negative in both crosses, but not significant in the

3D7x3D1 cross (Tables S2 and S3, 1A5x1E4 correlation = -0.44 and 3D7x31 correlation = -

0.12). As lower mean grey values correspond to higher amounts of melanin, these

correlations indicate that isolates that produce more melanin in a benign environment were

more tolerant to KCl stress, although this relationship was not significant at 8 dpi in the

3D7x3D1 cross.

3.3. QTL results

We identified at least one significant QTL for all but three traits (KCl melanisation rate, KCl

mean grey value at 12 dpi) in the 1A5x1E4 cross and all but one trait (KCl-tolerance affecting

melanisation rate) in the 3D7x3D1 cross (Table 1, for plots of all QTLs see Supplementary

Material S2 and S3). In general, the traits lacking significant QTLs had relatively low

heritability (Table 1) and thus the power to detect a QTL for those traits was relatively low.

All QTL positions, interval sizes and numbers of genes within the interval are presented in

Tables 2 and 3. Overlapping QTLs were grouped and given a unique QTL identifier. As

expected, traits that were highly correlated and/or statistically coupled often had a QTL

peak at the same, or overlapping positions (e.g. the 1A5x1E4 grey values in the control

environment at 8 dpi and 12 dpi mapped to the same position on Chr 2: 1438934bp; the

3D7x3D1 QTL for radius at 12 dpi and growth rate in the control environment mapped to

Chr 8: 1100862bp).

3.4. QTLs shared across environments, traits, and time points

Several QTLs were shared amongst benign and salt-stress environments, across time points

and traits. Only one of these shared QTLs is novel to this study, the other shared QTLs were

found in earlier studies investigating different stressors. The single novel QTL shared

between benign and salt-stress environments was found in the 3D7x3D1 cross on Chr 8

(1100.86-1221.33 Kb, QTL No. 9) and explained variation in seven traits including growth in

the control and KCl environments, as well as mean grey value in the control environment

and KCl tolerance at 8 dpi (Figure 1, Table 2).

Of the QTLs shared across benign and KCl environments that were identified here and found

in earlier studies using different stresses (e.g. cold, fungicide), the most striking – in terms of

the number of traits and the magnitude of the LODs, was the QTL on Chr 11 between

507.36-658.11 Kb (QTL No. 12) in the 3D7x3D1 cross. This QTL explained variation in 11 of

the 18 traits, and in both environments (Table 2). This QTL was identified previously and was

found to explain variation in growth, fungicide sensitivity and melanisation in this cross

(Lendenmann et al., 2015, 2014). The next largest, in terms of number of traits, was a QTL in

the 1A5x1E4 cross on Chr 8 (274.792- 346.224 Kb) that explained variation in growth rate in

both control and KCl environments and mean grey value in the control environment (Figure

2, Table 3). This QTL on Chr 8 was also found to explain variation in response to oxidative

stress (Zhong et al., 2021). Another notable QTL in terms of high LOD (>10) was on Chr 10

(775.47 – 924.42 Kb, QTL No. 10) in the 3D7x3D1 cross, which explained variation in

melanisation rate and colony radius in the control (Table 2). This QTL was previously found

to explain variation in growth-related traits at 10^{0}C (Lendenmann et al., 2014) and under

oxidative stress (Zhong et al., 2021).

3.5. Unique QTLs associated with the KCl environment

Several novel QTLs that were unique to the KCl environment were also found (Table 2, 3). In

the 3D7x3D1 cross, two QTLs were found for traits related to colony radius: the first on Chr

3 (QTL No. 3) explained variation in colony radius at 8 dpi and KCl-tolerance affecting colony

radius at 8 and 12 dpi; and the second on Chr 4 (QTL No. 5) explained variation in colony

radius at 8 dpi and KCl-tolerance affecting colony radius at 8 dpi. For melanin-related traits,

four QTLs were found in the 3D7x3D1 cross: the first on Chr 1 (QTL No. 1) explained

variation in three traits (melanisation rate, KCl-tolerance affecting melanisation rate and

grey value at 12 dpi); the second on Chr 2 (QTL No. 2) explained variation in grey value at 12

dpi; the third on Chr 6 (QTL No. 6) explained variation in melanisation rate; and the fourth

on Chr 8 (QTL No. 8) explained variation in grey value at 8 dpi. In the 1A5x1E4 cross a total

of six QTLs related to colony radius were found. The first three were on Chr 1: QTL No. 2

explained variation in KCl-tolerance affecting colony radius at 12 dpi; QTL No. 3 explained

variation in KCl-tolerance affecting growth rate; and QTL No. 4 explained variation in colony

radius at 12 dpi. The fourth unique KCl environment QTL Chr 5 (QTL No. 5) explained

variation in KCl-tolerance affecting growth rate; the fifth on Chr 6 (QTL No. 11) explained

variation in radius at 8 dpi; and the sixth on Chr 8 (QTL No. 14) explained variation in KCl-

tolerance affecting colony radius at 8 dpi. One QTL unique to the KCl environment was

found on Chr 8 (QTL No. 15) that explained variation in both radius- and melanin-related

traits, specifically KCl-tolerance affecting colony radius and grey value at 8 dpi.

To identify candidate genes associated with KCl stress, we focused attention on novel QTLs

with a high LOD (>10), relatively narrow intervals, and that explained a relatively high

proportion of the variation for a KCl-associated trait. Only QTL No. 3 in the 3D7x3D1 cross

met these requirements (Table 2). This QTL had a high LOD (>10) for colony radius at 8 dpi in

the KCl environment and KCl-tolerance affecting the radius at 8 dpi. The LOD peaked at 22.8

at chromosome position 737773 bp. This QTL explained 33% of the phenotypic variance in

colony radius at 8 dpi and 23% of the variance in KCl-tolerance affecting the radius at 8 dpi.

Isolates that had the non-reference allele at the LOD peak marker (X3_737773) were smaller

(F=113.1, df=1,233, p=<0.001, Figure 3c) and less KCl-tolerance affecting the radius at 8 dpi

(F=113.1, df=1,233, p=<0.001, Figure 3d). The Bayes confidence interval of this QTL spanned

163 Kb and contained 105 SNPs and 36 genes (Table 2, Figure 3b). For this interval we

present data on the genes in the reference genome (3D7), the orthologous genes in 3D1,

the gene sequence similarity (including ±200bp of flanking regions) between the parents,

the gene's putative encoded function (PFAM database) and if there is any evidence of

differential gene expression between the parents across all available datasets (Table 4 and

see Table S4 for full details).

3.6. Candidate genes within QTL No. 3 in 3D7x3D1

Several genes had SNP variation predicted to have a high impact based on SNPEff analysis

(Table S5, *ZT3D7_G3622, ZT3D7_G3623, ZT3D7_G3624, ZT3D7_G3632, ZT3D7_G3633,*

ZT3D7_G3635, ZT3D7_G3649). A transposable element (6927bp) was found within 200bp of

ZT3D7_G3623, and one gene, *ZT3D7_G3649* is a singleton – with no orthologue present in

any of the other 18 *Z. tritici* reference genome sequences (Table S4). Based on predicted

protein function, and sequence and gene expression variation between the parents, we

consider six genes to be the most promising candidates for further analyses: *ZT3D7_G3622*,

ZT3D7_G3635, *ZT3D7_G3637*, *ZT3D7_G3639*, *ZT3D7_G3655* and *ZT3D7_G3658* (this order

reflects their relative positions on chromosome 3).

ZT3D7_G3622 has a loss of function SNP (Chr 3:598092) in 3D1 that is predicted to cause a

loss of a predicted start codon. The gene contains a single exon, encodes a protein length of

550 amino acids (aa) and was predicted to contain a fungal-specific transcription factor

domain and Zn(2)-C6 fungal-type DNA-binding domain (Table 4). The orthologue in the 3D1

genome is *ZT3D1_G3567*. It encodes a longer protein (722 aa) and has two exons, the first

exon's product encodes a Zn(2)-C6 fungal-type DNA-binding domain and the second exon's

product is the same as that of *ZT3D7_G3622*. The predicted protein in 3D1 is the same as

the orthologue in IPO323 (*Mycgr3T108446*). Although we do have reads mapping to this

region, and when grown in vitro on MM this gene is downregulated in 3D1 compared to 3D7

(Table S4), our RNA sequence data does not clearly support gene models in either 3D1 or

3D7, there are no clear exon/intron boundaries. This suggests that the annotation of this

gene may require manual curation.

ZT3D7_G3635 has a high impact SNP variant (Chr 3:681906) that is predicted to cause a

frameshift. The gene has 6 exons in 3D7 and encodes a protein length of 1078 aa. Based on

INTERPROSCAN and Ensemble Fungi, the protein has multiple predicted domains; the

largest is a protein kinase domain. The protein also contains a Serine/threonine-protein

kinase active site and the molecular function of the first GO term (GO:0004674) is protein

serine/threonine kinase activity – suggesting that this gene is likely to be involved in osmotic

stress response (Hohmann, 2002). The RNA sequence data supports the gene model. The

orthologue in 3D1 is *ZT3D1_G3580*, which has 3 exons and multiple predicted domains that

do not differ from those predicted for *ZT3D7_G3635*. The gene is up-regulated in vitro on

YSB and down-regulated in planta at 28d in 3D1 compared to 3D7 (Table S4).

ZT3D7_G3637 has one exon, encodes a protein length of 1285 aa, and includes multiple

predicted domains including a DEAD/DEAH box helicase, which is involved in plant

responses to osmotic stress (Nguyen et al., 2018). The orthologue in 3D1 (*ZT3D1_G3583*)

encodes a similar protein as *ZT3D7_G3637* – has the same length and has the same

predicted domains. There are no high impact SNPs between the parents, but several

moderate and low impact variants were identified (Table S5). The RNA sequence data

supports the annotation, and the gene is down regulated in 3D1 in both YSB and MM (Table

S4).

ZT3D7_G3639 has 4 exons in 3D7. The gene sequence similarity between 3D7 and 3D1 is

98.66% and there is an 8bp intronic deletion in 3D1. This gene belongs to the Major

Facilitator Superfamily (MFS) and encodes a protein length of 512 aa. According to the

Phobius analysis, the gene contains 12 transmembrane segments (Table S4). Proteins with

transmembrane activity and transmembrane segments are likely to play an important role

in sensing and regulating osmotic stress (Hohmann, 2002). The orthologue in 3D1 is

ZT3D1_G3580, which has the same number of exons, encodes the same protein length and

has the same predicted function. The RNA sequence data supports the gene model, and the

differential expression analysis found the gene is downregulated in planta at 12 and 14 days

in 3D1 compared to 3D7 (Table S4).

ZT3D7_G3655 has one exon and encodes a protein length of 848 aa. The gene contains 10

transmembrane segments (Table S4) and encodes a voltage-gated chloride channel domain

– thus it is likely to be involved in osmotic stress responses. The gene sequence between the

parents is almost identical, but there is one moderate effect SNP within the gene (Table S5).

When grown in vitro on MM this gene is upregulated in 3D1 compared to 3D7. The RNA

sequence data supports the gene model annotation.

ZT3D7_G3658 has four exons and encodes a protein length of 614 aa. Its orthologue in 3D1

(*ZT3D1_G3604*) has three exons and encodes 572 aa. The gene encodes a vacuolar protein

sorting-associated protein 17, Vps17. Vacuoles play an important role in osmotic stress

responses: their fission and fusion enable cells to maintain turgor pressure and regulate cell

volume (Richards et al., 2010). Comparison of the 3D1 gene sequence to the 3D1 genome

using BLAST revealed that the 3D7 gene sequence was identical, but the 3D1 orthologue

annotation is shorter and has only three exons. Our RNA sequence data supports the 3D1

annotation with only 3 exons, as there were no RNA sequence reads mapping to exon 1 of

the *ZT3D7_G36580*. This gene was downregulated in vitro in both MM and YSB in 3D1

compared to 3D7.

4. Discussion

We were able to map major effect loci affecting colony growth and melanisation in the

presence of potassium chloride in both crosses. We focussed on QTLs that were unique to

the KCl environment to identify candidate genes that are specific to osmotic stress – as

opposed to finding genes that may be more generally related to intrinsic growth or melanin

production. A QTL on Chr 3 in the 3D7x3D1 cross was the most promising for identifying

genes specifically associated with KCl stress. We highlighted the six most promising

candidate genes in this QTL based on their predicted encoded functions, sequence similarity

and transcriptional profiles. We also identified several QTLs shared across environments and

confirmed that several previously identified QTLs, which explained variation in growth and

melanisation across different environmental stressors (Table 2 and 3, e.g. 3D7x3D1 Chr 11

and 1A5x1E4 Chr 8), also explained trait variation in the presence of osmotic stress.

4.1. KCl tolerance

In the presence of potassium chloride at 0.75M, colonies were typically smaller and less

melanised compared to the control environment (Table S1). A reduction in colony radius in

the KCl environment supports our hypothesis that this environment was stressful to *Z. tritici*.

Our results suggest that growth rate in the 1A5x1E4 cross was less affected by KCl-induced

osmotic stress, as mean growth rate was not reduced in the KCl environment, compared to

the 3D7x3D1 cross that exhibited a significantly reduced growth rate (Table S1). However,

we note that the overall difference in growth rate in the 3D7x3D1 progeny between the

control and KCl environments was relatively small (~6% reduction in growth, Table S1).

When experiencing osmotic stress, an increase in melanin is thought to protect cells by

strengthening the fungal cell wall, but a reduced metabolic capacity under stress may limit

the ability of a cell to produce melanin. The mean colony grey values were higher, indicative

of lower levels of melanin, for strains grown in the KCl environment (Table S1). This suggests

that the *Z. tritici* response to osmotic stress did not involve an increased production of

melanin to protect the cells as was observed in halotolerant species (Kogej et al., 2007). We

did observe however, that isolates that were more melanised in the benign control

environment were more tolerant to salt stress – their growth rates were higher and their

reductions in melanisation were lower compared to isolates that were less melanised in the

control environment (Table S2, S3). This suggests that strains having a greater intrinsic

capacity to become melanised are more tolerant to salt stress.

4.2. Shared QTLs

Some QTLs that explained variation in growth and melanisation in the stressful environment

overlapped with QTLs for trait variation in the control environment. Prominent examples

include the Chr 11 QTL at position 616039 in the 3D7x3D1 cross and the Chr 8 QTL at

position 274792- 346224 in the 1A5x1E4 cross. These genomic regions may harbour genes

that affect intrinsic growth of colonies under a wide range of environments and/or genes

with pleiotropic effects. Many of the QTLs shared across environments were identified in

earlier experiments and discussed previously (Lendenmann et al., 2015; Zhong et al., 2021).

The QTL on Chr 11 at position 616039 was shown to be due to differential expression of the

Zmr1 gene encoding a transcription factor that regulates synthesis of DHN melanin in *Z.*

tritici (Krishnan et al. 2018). *Zmr1* was shown to have pleiotropic effects on melanin

production, colony growth rates and fungicide resistance (Krishnan et al. 2018).

4.3. QTLs uniquely associated with the KCl environment

This QTL analysis enabled us to identity large effect loci that explain variation in traits under

KCl stress. Among the QTLs uniquely associated with KCl stress, most had moderate LODs

with each QTL explaining only a small proportion of the phenotypic variance (3D7x3D1

maximum = 0.10, 1A5x1E4 maximum = 0.11). The exception was QTL No. 3 in the 3D7x3D1

cross, which explained up to 33% of the phenotypic variance. This QTL on chromosome 3

was associated with differences in radius at 8 dpi and KCl-tolerance affecting radius at 8 and

12 dpi. We targeted this QTL for further investigation seeking to identify genes specifically

associated with KCl tolerance. It is interesting to note that no LOD peak was detected for

colony radius at 12 dpi, even though radius at 12 dpi was correlated with radius at 8 dpi

(Table S2, correlation = 0.58, p<0.001). As growth rate is based on the change in radius

between 8 and 12 dpi, we might expect growth rate in KCl to share the QTL for colony size

at 8 dpi. To better understand why growth rate does not share this QTL, we examined the

trait correlations. Growth rate is highly correlated with colony radius at 12 dpi (0.93) and

less strongly correlated with colony radius at 8 dpi (0.26) (Table S2). There is less variation in

colony radius at 8 dpi, thus colony radius at 12 dpi has a greater impact on the growth rate.

The QTL also explained variation in KCl-tolerance affecting the radius at 8 and 12 dpi,

although the association with KCl-tolerance affecting the radius at 12 dpi explained

relatively little phenotypic variation (0.08). Thus, the QTL peak is more strongly associated

with radius and tolerance differences at an earlier stage of colony growth. A similar pattern

was observed in the chromosome 11 QTL associated with melanisation (Krishnan et al.,

2018; Lendenmann et al., 2014), with the QTL LOD score very high in the younger colonies

and decreasing as colonies aged. After cloning and characterizing the *Zmr1* gene and its

functions, this pattern could be explained based on the finding that the difference in *Zmr1*

expression (and the resulting difference in melanisation) was significantly higher in younger

colonies than in older colonies (Krishnan et al., 2018). We hypothesize that QTL no. 3

associated with KCl tolerance is underpinned by a similar mechanism that is specific to the

earlier stage of colony growth.

4.4. Candidate genes in the QTL for radius and KCl-tolerance in radius at 8 dpi in the

3D7x3D1 cross

Among the 36 genes found in the interval spanning this QTL on chromosome 3, we consider

six genes to be the most promising candidates to be involved in sensing and responding to

salt stress and responsible for this QTL. Below we consider each in turn in the order of their

relative positions on chromosome 3.

The first candidate gene, *ZT3D7_G3622*, is predicted to encode a fungal-specific

transcription factor. The orthologue in the IPO323 reference genome (assembly MYCGR

v2.0) is *Mycgr3T108446* and it has 21 paralogs. Transcription factors (TFs) regulate gene

expression and are important regulators of molecular responses to environmental stimuli

(e.g. Hohmann, 2002; Shi et al., 2021), making this gene a good candidate. The *Zmr1* gene

identified in this same cross and shown to be responsible for differences in melanisation,

with greater effects in younger colonies, is a transcription factor (Krishnan et al. 2018). The

SNP variant in 3D1 (Chr 3:598092) is predicted to cause a loss of the start codon that is likely

to cause a loss of function. In line with this, the mean radius of isolates carrying the 3D1

allele was smaller than the isolates carrying the 3D7 allele. In *F. graminearum* mutants

lacking a similar TF (called *Fss1*) were sensitive to high sodium and lithium, but they were

not sensitive to elevated potassium (Son et al., 2015).

The second candidate, *ZT3D7_G3635* is predicted to encode a serine/threonine-protein

kinase active site, which is known to regulate MAP kinases (Hohmann, 2002; Li et al., 2019).

A SNP variant found in the 3D1 allele is predicted to cause a frameshift that is likely to cause

a loss of function. The ortholog in the IPO323 genome is *Mycgr3G56270* and it has one

paralog in the reference genome. A serine/threonine kinase from *Aspergillus flavus*,

AflSte20, was recently functionally validated using a knockout mutant. This study showed

that the knockout had higher sensitivity to osmotic stress compared to the wild type (Li et

al., 2019), confirming the role of this serine/threonine-protein kinase in osmotic stress

response in a different fungal plant pathogen.

The third candidate, *ZT3D7_G3637* is predicted to encode a DEAD/DEAH box helicase

domain. The reference genome orthologue is *Mycgr3G99415* and it has 6 paralogs. The

DEAD – box helicases are the largest RNA helicase (RH) subfamily. In plants they play

multiple functional roles and are important in stress responses (Pandey et al., 2020). For

example, in *Arabidopsis thaliana AtRH17* is involved in salt-stress tolerance (Nguyen et al.,

2018), in rice *OsABP* is upregulated in response to salt (Macovei et al., 2012) and in barley

HvD1 accumulated in salt- and cold-stressed plants (Nakamura et al., 2004). In wheat, the

DEAD-box helicase *TaRH1* that was first identified as a component of the defence response

to stripe rust (*Puccinia striiformis f.* sp. *tritici*) was also found to be involved in plant

response to multiple stressors, including salt stress (Zhang et al., 2014). SNP variation found

between the parents was predicted to have only moderate effects, no high impact SNPs

were detected. We chose this gene as a promising candidate based mainly on its known

associations with salt stress in plants and the significant differences in expression found

between 3D1 and 3D7 while growing in vitro.

A gene encoding a Major Facilitator Superfamily (MFS) domain, *ZT3D7_G3639* is the fourth

candidate we highlight. The reference genome orthologue is *Mycgr3G85019* and it has 58

paralogs. MFS transporters facilitate the movement of small solutes across cell membranes

in response to chemiosmotic gradients – making them likely to play an important role in

sensing and regulating osmotic stress (Hohmann, 2002). We chose this gene as a promising

candidate based mainly on its known functions and associations with multiple stressors

including fungicides (Dos Santos et al., 2014; Omrane et al., 2015).

The fifth highlighted gene is *ZT3D7_G3655* that encodes a voltage gated chloride channel

domain and membrane transport domains. The reference genome orthologue is

Mycgr3T20761 and there are three paralogues in the IPO323 reference genome. Chloride

channels are important in animal cell volume regulation (Jentsch et al., 2002), fungal

oxidative stress response (Oddon et al., 2007), fungal pathogen virulence (Cañero and

Roncero, 2008; Zhu and Williamson, 2003) and plant immunity (Guo et al., 2014). The role of

chloride channels in response to salt stress is well established in animals (Jentsch et al.,

2002), and to a lesser degree in plants (Barbier-Brygoo et al., 2011), but less is known from

fungi. For one of the better-known families of chloride channels, the CLC family, a single

gene (*GEF1*) has been identified in yeast. In two pathogenic fungi (*Cryptococcus neoformans*

and *Fusarium oxysporum*) – the GEF1 homologue has been shown to influence virulence

(Cañero and Roncero, 2008; Zhu and Williamson, 2003). The gene sequence between the

parents is highly conserved, however gene expression differences were observed in minimal

media, so differences may exist in the regulatory sequence.

The final candidate we highlight is *ZT3D7_G3658*, which encodes a Vacuolar associated

protein 17, *Vps17*. The orthologous gene in the reference genome is *Mycgr3T108464* and

there are six paralogues. Comparing the DNA sequence of this gene against the 3D1 genome

revealed an identical sequence, but the annotation of the 3D7 and 3D1 differs – the 3D1

orthologue protein length is shorter and the gene has one fewer exons. Our RNA sequence

data suggests that the *ZT3D7_G3658* annotation may be incorrect - there were no reads

mapping to the first, small exon of this gene. This suggests that the annotation of this gene

may need manual curation before conducting further analyses. The gene was

downregulated in YSB and MM in 3D1 compared to 3D7, demonstrating that transcription

of this gene varies between the parents. Vacuoles play important roles in both osmotic

stress responses (Richards et al., 2010) and virulence (Li et al., 2021). Under salt stress the

vacuoles can undergo fusion and fission – changing in size and number (Richards et al.,

2010). *Vsp17* knockouts in *S. cerevisiae* had many small vacuoles but were resistant to

osmotic stress and could sporulate (Köhrer and Emr, 1993). We are unaware of any work

that has investigated the role of *Vsp17* in salt-stress responses of filamentous fungi.

4.5. QTL reproducibility

This dataset allowed us to explore the repeatability of previously identified QTLs from other

published datasets. One earlier dataset used the same mapping populations but different

experimental conditions and treatments, with a different reference genome and different

genetic markers and map information (Lendenmann et al., 2015, 2014). A second earlier

dataset used the same isolates and experimental conditions, but different treatments and

different marker and map information (Zhong et al., 2021). Comparisons across these three

data sets yielded several consistent QTLs. Our analyses confirmed that a region on

chromosome 11 in the 3D7x3D1 cross (Table 2, QTL No. 11) is associated with variation in

melanisation and growth in multiple environments (e.g. under fungicide stress in

Lendenmann et al., 2015; under oxidative stress in Zhong et al., 2021). This QTL contains the

Zmr1 transcription factor that was shown to regulate melanin production (Krishnan et al.,

2018). We hypothesize that *Zmr1* also explains the variation we observed under salt-stress.

A QTL associated with multiple growth and melanin traits in the 1A5x1E4 cross on

chromosome 8 (Table 3, QTL No. 13) overlaps with a previously identified QTL on

chromosome 8 that was associated with colony growth and tolerance to oxidative stress

(Zhong et al., 2021). These findings illustrate that many of our QTLs are reproducible across

experiments and increases our confidence that we have identified biologically meaningful

genomic regions using our automated image analyses.

CRediT authorship contribution statement

Jessica Stapley: Formal analysis, Data curation, Writing - Original Draft, Writing - Review &

Editing, Resources, Data curation. Bruce A. McDonald: Conceptualization, Methodology,

Writing - Review & Editing, Supervision, Project administration, Funding acquisition.

Declaration of Competing Interest

The authors declare that they have no known competing financial interests or personal

relationships that could have appeared to influence the work reported in this paper

Data availability

Phenotypic and SNP data are available at ETH Data repository (DOI: 10.3929/ethz-b-

000550424). All code and details of the analysis are available at

(https://github.com/jessstapley/QTL-mapping-Z.-tritici).

Acknowledgments

We thank Ziming Zhong, Tiziana Valeria Vonlanthen, Bethan Turnbull, Susanne Dora, Sarah

Furler, Alexandra Waltenspühl, and Jasmin Wiedmer for collecting the phenotypic data and

Javier Palma-Guerrero for supervision. We thank Marcello Zala for assistance in the

laboratory. This research was supported by the Swiss National Science Foundation

(31003A_155955 granted to BAM).

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Figure 1. Genome scan from the 3D7x3D1 cross of a) growth rate and b) melanisation rate

in the control (grey) and in the KCl environment (orange). The dashed horizontal lines

indicate the LOD significance thresholds for each trait based on permutation.

Figure 2. Genome scan from the 1A5x1E4 cross of a) growth rate and b) melanisation rate in

the control (grey) and in the KCl environment (orange). The dashed horizontal lines indicate

the LOD significance thresholds for each trait based on permutation.

Figure 3. QTL interval on Chromosome 3 for colony radius at 8 dpi in the 3D7x3D1 cross. a)

LOD peak and Bayes credible interval (shaded) with SNP positions indicated on the x-axis as

black vertical lines, b) Alignment of the genes within the QTL interval between the 3D7 and

3D1 parental strains – orange-coloured genes are the six most promising candidate genes

described in the main text and c) allelic effect of the SNP with the highest LOD on colony

radius at 8 dpi and d) allelic effect of the SNP with the highest LOD on KCl-tolerance

affecting colony radius at 8 dpi.

Supplementary Material

S1. Summary of genetic linkage maps.

S2. Genome scans from the 3D7x3D1 cross for all traits.

S3. Genome scans from the 1A5x1E4 cross for all traits.

Table 1. Summary of the QTL mapping results for all traits. The details of each trait are as follows, data type indicates if the measurement was
made at a single time point (days post inoculation, DPI) or if it was calculated across two time points (Rate) or across two environments
(Tolerance) (see Methods for details). The environment indicates the treatment conditions the colonies were grown in, control, potassium
chloride (KCl), or calculated as a ratio across both environments (both). The measurement column indicates if the measurement was colony
radius, grey value (indicator of melanin) or the change in radius or melanin over time (growth rate, melanisation rate). For example, row 1:
colony radius measured at 8 DPI in the control environment. The QTL peak columns show the chromosomal position of significant QTL peak for
each trait in the 3D7x3D1 and 1A5x1E4 crosses respectively, and the final column contained estimated heritability (h^2) for each trait.
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Data Type	Environment	DPI	Measurement	QTL peak 3D7x3D1	h^2	QTL peak 1A5x1E4	h^2
Time point	Control	8	Radius	3,8,10,11	0.53	2,3,8	0.56
Time point	Control	12	Radius	3,7,8,10,11	0.62	7,8	0.48
Time point	KCl	8	Radius	3,4	0.64	6,	0.50
Time point	KCl	12	Radius	3,8	0.60	1,8	0.52
Time point	Control	8	Grey value	8,10,11	0.42	1,2,3,8	0.63
Time point	Control	12	Grey value	11	0.22	2,3,8	0.44
Time point	KCl	8	Grey value	8,11	0.54	8	0.34
Time point	KCl	12	Grey value	1,2,11	0.34	No Sig. QTL	0.29
Rate	Control	8-12	Growth rate	3,8,11	0.61	7,8	0.33
Rate	KCl	8-12	Growth rate	3,8,11	0.45	8,	0.49
Rate	Control	8-12	Melanisation rate	10	0.28	2,3,12	0.54
Rate	KCl	8-12	Melanisation rate	1,6	0.32	No Sig. QTL	0.27
Tolerance	Both	8	Grey value	8,10	0.37	1,2,3,8	0.58
Tolerance	Both	12	Grey value	11	0.18	No Sig. QTL	0.34
Tolerance	Both	8	Radius	3,4,10	0.59	8	0.42
Tolerance	Both	12	Radius	3,10,11	0.47	1	0.37
Tolerance	Both	8-12	Growth rate	11	0.40	1,5	0.45
Tolerance	Both	8-12	Melanisation rate	No Sig. QTL	0.14	1	0.32

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810　**Table 2.** Description of all QTLs identified from the 3D7x3D1 cross, including map and physical position of the LOD peak marker, the bayes
811　credible interval length and the number of genes within the interval. PVE is the proportion of variance explained by the QTL. QTL number is a
812　number assigned to a particular QTL peak which identifies overlapping QTLs.
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Type	Measurement	Environment	DPI	Chr	Pos (cM)	LOD	Pos (bp)	No. markers	Interval Start (bp)	Interval end (bp)	Interval size (bp)	No. Genes	PVE	QTL number
Rate	Melanisation rate	KCl	8-12	1	457.43	6.18	1590235	2915	1035571	2769729	1734.158	598	0.10	1
Tolerance	Melanisation rate	Ratio	8-12	1	540.00	3.40	2295356	9000	788430	6389989	5601.559	1977	0.06	1
Time point	Grey value	KCl	12	1	576.00	5.99	2717234	519	2295128	2769729	474.601	174	0.10	1
Time point	Grey value	KCl	12	2	261.00	5.07	691536	141	679366	711981	32.615	15	0.09	2
Tolerance	Radius	Ratio	12	3	142.92	4.49	719549	553	396401	944712	548.311	125	0.08	3
Tolerance	Radius	Ratio	8	3	144.00	14.62	737773	105	588483	751983	163.5	36	0.23	3
Time point	Radius	KCl	8	3	144.15	22.83	737773	94	588483	737773	149.29	31	0.33	3
Rate	Growth rate	Control	8-12	3	415.00	5.31	2537296	2204	1982744	3347765	1365.021	462	0.09	4
Time point	Radius	Control	12	3	456.63	8.11	2993469	427	2971359	3196103	224.744	79	0.13	4
Time point	Radius	Control	8	3	482.67	7.09	3056536	428	2982086	3199430	217.344	76	0.12	4
Time point	Radius	KCl	12	3	485.65	4.99	3110053	3084	516017	3203947	2687.93	813	0.08	4
Rate	Growth rate	KCl	8-12	3	485.65	4.93	3110053	845	2971359	3347765	376.406	132	0.08	4
Tolerance	Radius	Ratio	8	4	400.00	4.28	1936846	1000	1845763	2448166	602.403	200	0.07	5
Time point	Radius	KCl	8	4	403.76	4.37	2013908	986	1867509	2448166	580.657	194	0.07	5
Rate	Melanisation rate	KCl	8-12	6	220.14	6.24	1767543	479	1568163	1981193	413.03	124	0.10	6
Time point	Radius	Control	12	7	186.78	4.47	1109088	1366	511101	1377451	866.35	265	0.08	7
Time point	Grey value	KCl	8	8	680.87	5.69	943669	732	502305	974947	472.642	123	0.10	8
Time point	Radius	Control	12	8	746.72	10.77	1100862	274	1091579	1164224	72.645	29	0.17	9
Rate	Growth rate	Control	8-12	8	746.72	7.31	1100862	542	1071962	1221280	149.318	65	0.12	9
Rate	Growth rate	KCl	8-12	8	747.72	9.66	1100862	667	1091692	1284370	192.678	77	0.16	9
Time point	Radius	KCl	12	8	751.61	9.79	1112633	670	1100862	1292397	191.535	76	0.16	9
Time point	Radius	Control	8	8	765.72	8.04	1221337	751	1091475	1354349	262.874	108	0.13	9

Tolerance	Grey value	Ratio	8	8	776.11	4.22	1350673	2608	962119	2165894	1203.775	457	0.07	9
Time point	Grey value	Control	8	8	776.11	4.35	1350706	2549	981398	2175005	1193.607	450	0.07	9
Time point	Grey value	Control	8	10	199.98	12.37	775475	160	769669	861699	92.03	32	0.20	10
Tolerance	Grey value	Ratio	8	10	199.98	10.68	775475	281	769669	919978	150.309	56	0.17	10
Rate	Melanisation rate	KCl	8-12	10	200.35	8.66	792011	299	769669	924425	154.756	59	0.14	10
Time point	Radius	Control	8	10	200.72	12.32	792272	299	769669	924425	154.756	59	0.19	10
Tolerance	Radius	Ratio	8	10	211.51	10.58	924412	283	792019	924425	132.406	53	0.17	10
Tolerance	Radius	Ratio	12	10	211.51	6.68	924414	1081	749521	1302645	553.124	211	0.11	10
Time point	Radius	Control	12	10	212.61	4.92	924425	540	749521	1054575	305.054	116	0.08	10
Time point	Radius	Control	8	11	183.07	4.20	507361	528	311631	698718	387.087	128	0.07	11
Tolerance	Radius	Ratio	12	11	188.28	5.60	581439	1340	74200	890544	816.344	268	0.09	11
Time point	Grey value	KCl	12	11	192.28	6.87	605810	408	463714	776764	313.05	99	0.11	11
Time point	Radius	Control	12	11	193.00	16.79	616039	25	601600	658119	56.519	16	0.26	11
Rate	Growth rate	KCl	8-12	11	193.00	5.47	616039	219	468760	698718	229.958	72	0.09	11
Rate	Growth rate	Control	8-12	11	193.00	19.63	616039	5	616039	658119	42.08	13	0.29	11
Time point	Grey value	Control	12	11	194.11	6.64	657530	498	514049	870059	356.01	121	0.11	11
Tolerance	Grey value	Ratio	12	11	194.11	5.55	657530	700	367519	870059	502.54	168	0.09	11
Time point	Grey value	Control	8	11	194.11	5.30	658070	244	463714	704331	240.617	77	0.09	11
Time point	Grey value	KCl	8	11	194.11	17.09	658119	5	616039	658119	42.08	13	0.26	11
Tolerance	Growth rate	Ratio	8-12	11	195.28	7.45	658119	370	500942	802650	301.708	98	0.12	11

Table 3. Description of all QTLs identified from the 1A5x1E4 cross, including map and physical position of the LOD peak marker, the bayes

credible interval length and the number of genes within the interval. PVE is the proportion of variance explained by the QTL. QTL number is a

number assigned to a particular QTL peak which identifies overlapping QTLs..

Type	Measurement	Environment	DPI	Chr	Pos (cM)	LOD	Pos (bp)	No. markers	Interval Start (bp)	Interval end (bp)	Interval size (bp)	No. Genes	PVE	QTL number
Time point	Grey value	Control	8	1	328.91	5.69	572065	402	442697	727696	284.999	92	0.10	1
Tolerance	Melanisation rate	Ratio	8-12	1	328.91	4.46	572065	463	461577	779936	318.359	102	0.08	1
Tolerance	Grey value	Ratio	8	1	328.91	5.01	572065	405	442697	732023	289.326	93	0.08	1
Tolerance	Radius	Ratio	12	1	559.12	4.07	1849200	4654	399579	5725682	5326.103	1964	0.07	2
Tolerance	Growth rate	Ratio	8-12	1	797.00	4.80	3405315	3068	399544	3488556	3089.012	1151	0.08	3
Time point	Radius	KCl	12	1	1104.00	4.03	5792644	4774	372886	5956797	5583.911	2056	0.07	4
Rate	Grey value	Control	8-12	2	70.00	6.56	574965	1211	405115	1914465	1509.35	509	0.11	5
Time point	Grey value	Control	12	2	204.66	4.55	1438934	618	1284536	2152463	867.927	281	0.08	6
Time point	Grey value	Control	8	2	204.66	18.08	1438934	185	1438863	1748215	309.352	95	0.27	6
Tolerance	Grey value	Ratio	8	2	204.66	15.97	1438934	184	1438934	1748215	309.281	95	0.24	6
Time point	Radius	Control	8	2	211.42	8.76	1631840	217	1438934	1776132	337.198	105	0.14	6
Tolerance	Grey value	Ratio	8	3	147.00	6.15	1325440	210	1243932	1684531	440.599	179	0.10	7
Time point	Grey value	Control	8	3	148.00	6.66	1325440	133	1325422	1592801	267.379	115	0.11	7
Time point	Radius	Control	8	3	163.74	5.83	1521535	725	355042	1592801	1237.759	374	0.10	8
Rate	Grey value	Control	8-12	3	193.38	11.99	1711037	290	1489966	2105601	615.635	180	0.19	9
Time point	Grey value	Control	12	3	198.73	4.04	1928783	860	1743111	2927356	1184.245	361	0.07	9
Tolerance	Growth rate	Ratio	8-12	5	399.00	4.34	2638913	155	2511016	2658480	147.464	48	0.07	10
Time point	Radius	KCl	8	6	179.00	4.48	1512508	594	778583	1557528	778.945	256	0.08	11
Rate	Growth rate	Control	8-12	7	158.82	4.43	539824	255	437699	651318	213.619	80	0.07	12
Time point	Radius	Control	12	7	158.82	4.67	539837	254	437718	651318	213.6	80	0.08	12
Rate	Growth rate	KCl	8-12	8	76.80	8.79	274792	82	256529	346224	89.695	25	0.14	13
Time point	Radius	KCl	12	8	76.80	8.94	274792	67	256529	313758	57.229	17	0.15	13

Time point	Grey value	Control	12	8	78.58	4.92	302772	882	89684	921404	831.72	308	0.08	13
Time point	Radius	Control	12	8	79.00	9.72	302772	80	260355	346224	85.869	24	0.16	13
Time point	Grey value	Control	8	8	79.00	12.12	302772	66	270834	346224	75.39	22	0.19	13
Time point	Radius	Control	8	8	79.00	11.06	302772	415	260101	732444	472.343	163	0.18	13
Tolerance	Grey value	Ratio	8	8	79.00	8.26	302772	339	89573	346224	256.651	90	0.14	13
Tolerance	Radius	Ratio	8	8	137.00	5.05	832180	1177	589084	1648406	1059.322	396	0.08	14
Time point	Grey value	KCl	8	8	164.00	4.33	1377926	1406	305075	1623437	1318.362	485	0.07	14
Rate	Grey value	Control	8-12	12	126.74	4.24	473688	275	441706	821603	379.897	101	0.07	15