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1 **Identification of genetic bases of male fertility in *Pyricularia oryzae* by Genome Wide Association**

2 **Study (GWAS).**

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16 **Abstract**

17 The reproductive system of an organism impacts the emergence and evolution of adaptive variants in
18 response to selective constraints. The understanding of the sexual mode of reproduction in pathogens
19 helps to understand their life history. In filamentous Ascomycete fungi, mating type system and the
20 production of gametes are required to reproduce sexually. In the phytopathogenic Ascomycete
21 *Pyricularia oryzae*, studies of the genetic determinants of sexual reproduction are still limited to the
22 mating type system. This study focuses on identifying the genes involved in male fertility through the
23 production of male gametes known as microconidia. We performed a GWAS analysis coupled with a
24 local score approach on a wild recombinant population of *Pyricularia oryzae* phenotyped for
25 microconidia production. We identified one genomic region significantly associated with the quantity
26 of microconidia produced. This region contained nine candidate genes, some of them annotated with
27 functions associated to sexual reproduction in model fungi such as *Neurospora crassa*, *Podospora*
28 *anserina* and *Sordaria macrospora*. The most promising candidate gene contains a Jumonji domain.
29 Proteins belonging to the Jumonji family are conserved among Eukaryotes and are known to be involved
30 in chromatin regulation.

31 **Introduction**

32 The potential of a pathogen population to adapt to its environment is driven by evolutionary
33 forces, among which recombination is key to reassort existing parental alleles and generate new
34 genotypes (McDonald and Linde, 2002). Sexual reproduction is the main mechanism producing
35 recombination at the genome scale. Thus, the generation of new genotypes thanks to sexual reproduction
36 in pathogen organisms likely contributes to its adaptation to the environment, to the host or to the
37 bypassing of host resistance. In addition to recombination, in several pathogen fungal species, sexual
38 reproduction may lead to the production of specific cells with enhanced survival in unfavorable
39 conditions or during dispersion.

40 In the heterothallic phytopathogenic fungus *Pyricularia oryzae* (syn. *Magnaporthe oryzae*), the
41 causal agent of blast disease on many cultivated crop with a major food interest (wheat, rice, maize...)
42 (Ou, 1985; Islam et al., 2016; Pordel et al., 2021), sexual reproduction has never been observed in the
43 field, whatever the host specificity of the population. However, evidences from biology, genetic and
44 genomic studies of populations isolated from rice suggest that sexual reproduction took place or is still
45 taking place in limited areas of the Himalaya foothills (the putative center of origin of *P. oryzae*; Zeigler,
46 1998), some populations present in this area exhibiting equilibrated frequencies of both mating types
47 and footprints of recombination (Saleh et al., 2012; Gladieux et al., 2018b; Thierry et al., 2022). These
48 populations belong to a single lineage. The three other genetic lineages that have spread worldwide are
49 clonal (Saleh et al., 2012; Gladieux et al., 2018b; Thierry et al., 2022) and show all genetic and biological
50 characteristics of asexual populations. Whether the loss of sexual reproduction is a cause or a
51 consequence of the spread of these lineages outside the center of origin is still a matter of debate. To
52 understand the evolution of the reproduction mode of rice blast populations, a preliminary step is to
53 better characterize the genetic determinants of fertility.

54 In filamentous fungi, the ability to reproduce sexually requires two conditions: the production
55 of functional sexual cells, i.e. male and female gametes, and the recognition, meeting and fusion between
56 these gametes. Recognition and encounter of a male gamete (either contained in antheridia, or present
57 as individualized specialized cells called spermatia) with a female gamete (the ascogonium) is the

58 starting point of the fertilization process. This process is governed by hormonal attraction, and is
59 controlled by a single mating-type locus with two alternative versions in heterothallic fungi (Metzenberg
60 and Glass, 1990). These idiomorphs comprise genes encoding proteins with an HMG-box DNA binding
61 motif that regulates the expression of genes involved in the pheromone/receptor machinery. The mating-
62 type system therefore rules the meeting and fusion of gametes (Debuchy et al., 2010). In *P. oryzae*, the
63 two idiomorphs of the *MATI* locus (*MATI.1* and *MATI.2*) correspond to 2 and 3 genes respectively, one
64 of them encoding a protein containing an alpha-box DNA binding domain (Kanamori et al., 2007).
65 Proteins encoded by HMG-box genes are transcription factors able to bind DNA, facilitate nucleoprotein
66 assembly (Giese et al., 1992), and their role in gamete production vary between species (Koopman,
67 2010). The mating-type system is not required for the production of gametes in *Podospora anserina*
68 (Coppin et al., 1993) or in *Neurospora crassa* (Ferreira et al., 1998). On the contrary to mating type
69 genes, the genetic determinants governing gamete production in Ascomycetes remain poorly
70 characterized. Some genes involved in spermatia production have been identified, such as *SsNsd* in
71 *Sclerotinia sclerotiorum* (Li et al., 2018) or H3K27 histone methyltransferase gene in *P. anserina*
72 (Carlier et al., 2021). These genes usually have pleiotropic effects. In *P. oryzae*, the loss of function of
73 a H3K4 histone methyltransferase is involved in pathogenicity and more generally in gene activation or
74 repression, but its involvement in male fertility was not tested (Pham et al., 2015). In *P. oryzae*, the
75 implication of the mating-type system in the production of male gametes was not assessed, and more
76 generally, the genetic determinants of this production are completely unknown. In this species,
77 specialized crescent-shaped cells called microconidia were recently shown to be the male fertilizing
78 elements (i.e. the spermatia), and male fertility was therefore defined as the ability to produce
79 microconidia (Lassagne et al., 2022).

80 Two complementary strategies have successfully been used to identify genes involved in fertility
81 in fungi: reverse and forward genetics. Apart from reverse genetics approaches based on transcriptomic
82 analyses (Garg and Jain, 2013; Riaño-Pachón et al., 2021; Strickler et al., 2012), reverse genetics
83 requires that genes governing the trait of interest have been characterized in one or several model species,
84 and therefore assumes that the genetic processes underlying the trait of interest are similar in the model

85 and focal species. Combined with comparative genomics, such candidate genes approaches brought
86 considerable insights in the understanding of sexual reproduction in Ascomycetes (Ellena et al., 2020;
87 Passer et al., 2022). However, such candidate-genes strategies cannot identify other uncharacterized
88 genes potentially involved in the phenotypic trait of interest. Alternatively, forward genetic approaches
89 allow detecting the genetic determinants of a given trait without a priori. Contrary to reverse approaches,
90 forward approaches aim to link genotypes to phenotypes by detecting statistical correlations between
91 genotypic markers and phenotypic observations in recombinant controlled or wild populations. These
92 approaches have long been restricted to Quantitative Trait Loci (QTL) analyses based on progeny of
93 controlled sexual crosses between parents with contrasted phenotypes, including in fungi (Foulongne-
94 Oriol, 2012). Genome Wide Association Studies (GWAS) approaches have later been used. GWAS was
95 initially developed to face the necessity of understanding human genetic diseases with no access to
96 controlled crosses (Hirschhorn and Daly, 2005). It relies on the detection of significant associations
97 between allelic frequencies at polymorphic positions (Single Nucleotide Polymorphisms, SNPs) along
98 the genome and phenotypic status for a given quantitative trait in wild recombinant populations. It
99 overcomes two major limitations of QTL approaches by avoiding the time-consuming production of
100 laboratory-controlled progenies and by giving access to many more generations of recombination.
101 Within the Fungal kingdom, GWAS proved its efficacy in several model and non-model species, notably
102 in understanding the genetic architecture of adaptation to the host (Dumartinet et al., 2022; Plissonneau
103 et al., 2017; Sánchez-Vallet et al., 2018), resistance to fungicides (Mohd-Assaad et al., 2016; Sanglard,
104 2019; Spanner et al., 2021), or communication between germinating neighbor conidia (Palma-Guerrero
105 et al., 2013). To our knowledge, fertility in fungi has never been investigated using GWAS. Here, we
106 used this approach to determine the genetic determinants of male fertility, specifically of the production
107 of microconidia in *P. oryzae*. In this purpose, we used full genome sequencing of strains from a wild
108 recombining population for which the production of microconidia was quantified.

109

110 **Material and Methods**

111 *Strains of Pyricularia oryzae*

112 Seventy-one strains of *P. oryzae* from a single population sampled in the locality of Yule in
113 Yunnan Province of China in 2008 and 2009 (Saleh et al., 2012) were characterized in this study. This
114 population, belonging to the recombinant lineage 1 described in Thierry et al. (2022), was shown to be
115 highly recombinant and suspected to be sexually reproducing (Saleh et al., 2012; Thierry et al., 2022).

116

117 *DNA extraction, sequencing, mapping and SNP calling*

118 DNA was extracted as described in Ali et al. (2023). Preparation of libraries and Illumina
119 NovaSeq 6000 sequencing was performed at Montpellier GenomiX (MGX), resulting in paired-end
120 reads of 150 bp. For each sequenced strain, genomic reads were mapped on GUY11 PacBio reference
121 genome with masked repeated elements (Bao et al., 2017; BioSample: SAMN06050153). We choose
122 GUY11 genome (assembly size 42.87 Mb, 56 contigs, contig N50=3.28 Mb) as a reference since this
123 strain belongs to the same genetic lineage as the strains studied here. Mappings were performed with
124 bwa mem v.0.7.17 and filtered with samtools v.1.10 (Danecek et al., 2021) for a mapping quality $q=20$
125 (Supplementary text 1: Script1). Mapping quality was assessed with samtools v.1.10 (Danecek et al.,
126 2021) and html-formatting of the report was performed with multiqc v.1.9 (Ewels et al., 2016;
127 Supplementary text 1: Script2). SNP calling was performed using bcftools v.1.10.2 (Danecek et al.,
128 2021) for a minimum depth of 10 reads per position and per individual, and a phred-scaled score above
129 30. SNPs were filtered for a Minimum Allele Frequency above 5%, and the insertions / deletions were
130 removed. Sites with more than 10% of missing data were deleted (Supplementary text 1: Script3).

131

132 *Population structure and LD decay*

133 The population structure of the 71 strains was inferred by a Principal Components Analysis
134 (PCA) using the R packages vcfR v.1.12.0 (Knaus and Grünwald, 2016) and adegenet v.2.1.7 (Jombart,
135 2008). Linkage disequilibrium (LD) decay was assessed with the PopLDdecay software (Zhang et al.,
136 2019), first by calculating pairwise r^2 between all pairs of SNPs with a maximum distance of 300 kb,
137 then by averaging all values in adjacent windows of 10 bp. Nucleotide diversity (P_i) was calculated with

138 EggLib v. 1.2 (De Mita and Siol, 2012) from the raw vcf file containing both invariant and variable sites
139 filtered for a maximum missing data of 10% and a minimum depth of 10 reads per position and per
140 individual.

141

142 *Microconidia production phenotyping*

143 Microconidia were produced following the protocol described in Lassagne et al. (2022). Briefly,
144 young mycelium grown on Rice Flour medium (20 g of rice flour, 15 g of Bacto agar, 2 g of Bacto Yeast
145 Extract in 1 L pure water with 500,000 U of penicillin G after autoclaving for 20 min at 120°C) was put
146 in 40 mL fresh homemade Potato Dextrose Broth (PDB). PDB was prepared with 200 g of sliced organic
147 potatoes boiled during 30 minutes in 800 mL pure water, filtrated through multi-layer gauze, completed
148 with 20 g glucose and replenished to 1 L. Liquid cultures were incubated during 3 days at 25°C then 6
149 days at 20°C with permanent shaking (150 rpm). After removing the mycelium by filtration on Mira cloth
150 filter film (22 µm), preparations were centrifuged at 4,500g for 15 min and microconidia were pelleted
151 and re-suspended in 1 mL sterile distilled water. The number of microconidia per mL was counted twice
152 with a Malassez cell under optic microscope (X40). Two cultures per strain, considered as independent
153 biological replicates, were carried out. The 71 strains were distributed in 9 batches corresponding to
154 different lots of fresh PDB and dates of experiment. To control for a possible batch effect, 11 strains
155 were duplicated in different batches.

156

157 *Statistical analysis of phenotypic values*

158 The microconidia production data were transformed in $\log(p+1)$ to reach normality of residues
159 and homoscedasticity. An analysis of variance (ANOVA) using a linear model was performed with the
160 lm function implemented in R. The model used was:

161

$$Y_{ijk} = \mu + a_i + b_j + c_k + E_{ijk}$$

162 where Y_{ijk} is the logarithm of the production of microconidia, μ is the intercept term, a_i the random
163 genotype effect of genotype i , b_j the replicate effect in replicate j , c_k the effect of batch k and E_{ijk} the
164 residuals term. This model was used to adjust the phenotypic values with the Least Square means
165 (LSmeans) procedure using the lsmeans R package (Lenth, 2016).

166

167 *Genome-wide association study and local score approach analyses*

168 GWAS was performed with GAPIT R Software package (version 3) (Wang and Zhang, 2021).
169 As genotypic information, we used the 29 largest scaffolds covering 99% of GUY11 genome. We tested
170 a Mixed Linear Model (MLM) and a Multi-Locus Mixed model (MLMM) (Segura et al., 2012)
171 accounting for genetic relatedness using kinship between strains, estimated using the centered identity-
172 by-step algorithm. The variant component procedure was used to estimate σ_a^2 and σ_e^2 using restricted
173 maximum likelihood, from the equation

174
$$\text{Var}(Y) = \sigma_a^2 K + R$$

175 where $\text{Var}(Y)$ is the phenotypic variance of the production of microconidia, σ_a^2 the genetic variance, K
176 the kinship matrix, and R the residual effect. Under the hypothesis of homogeneous variance, $R = I\sigma_e^2$,
177 where I is an identity matrix and σ_e^2 is the unknown residual variance. Broad sense heritability defined
178 as the proportion of genetic variance over the total variance was calculated as follows: $h^2 = \sigma_a^2 / (\sigma_a^2 +$
179 $\sigma_e^2)$. A p-value of genotype-phenotype association was considered significant when it was smaller than
180 the threshold calculated using the Dunn-Šidák method (Šidák, 1967), calculated as:

181
$$-\log_{10}[1-(1-\alpha)^{1/N}]$$

182 with α the Type I error (here 0.05) and N the number of SNP analyzed.

183 To get a better resolution on sub-significant peaks of p-values, we used the local score method
184 (Fariello et al., 2017). This method is designed to cumulate local association signals based on p-values
185 found with GWAS method, and is efficient in detecting loci with moderate to small effects (Bonhomme
186 et al., 2019). The aim of the local score method is to identify the genome segments that have a higher

187 density of SNPs with medium to high signal of association, compared with the rest of the genome. The
188 ξ parameter determines the range of p-values contributing to the local score: if p_i is the p-value of the i^{th}
189 locus, then the score is taken as $X_i = -\log_{10}(p_i) - \xi$, meaning that only p-values under $10^{-\xi}$ will contribute
190 positively to the score and p-values above will subtract from the signal. P-values distribution was
191 considered as uniform, and we retained $\xi=2$ and a Type I error risk of $\alpha=1\%$ as local score parameters.

192

193 *Analysis of genomic regions associated with microconidia production*

194 Local LD was assessed with the open source software LDBlockShow (Dong et al., 2021) in the
195 genomic regions significantly associated with microconidia production. To precisely identify the
196 candidate genes in these regions, we performed *de novo* gene prediction from the GUY11 assembly
197 using BRAKER1 and AUGUSTUS v3.0.3 software (Hoff et al., 2016), with the same RNA-sequencing
198 data for gene prediction as in Pordel et al. (2021). Gene models of GUY11 (including 2000 bp upstream
199 and downstream of the gene) were compared to their homologs in the 70-15 reference genome (Dean et
200 al., 2005). For the most significant candidate region, to overcome possible errors of annotation, gene
201 models were, when necessary, corrected for start, stop and introns positions with the 70-15 reference
202 genome or *de novo*, so that the coding sequences (CDS) correspond to functional (non-interrupted)
203 protein sequences. For each gene in this most significant candidate region, polymorphic sites in the 71
204 *P. oryzae* strains were then extracted from the vcf using the coordinates of the corrected gene model in
205 the GUY11 reference genome (Table 1).

206 To infer haplotype groups, we used Multiple Correspondence Analysis (MCA) implemented in
207 the FactoMineR package (Lê et al., 2008) using the SNP markers significantly associated to the
208 production of microconidia. We performed an ANOVA to test for the effect of the haplotype group on
209 the phenotypic values of the 71 strains, and used Tukey tests for pairwise comparisons of the mean
210 phenotypic value per haplotypic group.

211 **Results**

212 *Genetic diversity, population structure and recombination within the P. oryzae population*

213 The total number of reads per strain (in million) ranged from 10.9 to 85.9 with a mean of 20.9
214 and a median of 19.3 (i.e. total sequence per isolate ranging from 1.6 to 12.9 Gb with a mean of 3.1 and
215 a median of 2.9 Gb). Reads were mapped on the GUY11 reference genome with repeated elements
216 masked, and only reads with a mapping quality above 30 were retained. For all strains, at least 75% of
217 the reads mapped on the GUY11 genome, covering more than 80% of it, and leading to a mean mapping
218 depth from 36X to 278X per strain (median 63X; GUY11 genome size of 42.87 Mb according to Bao et
219 al., 2017; Supp. Figure 1). After filtration, 19,331 high-confidence SNPs distributed on the 56 scaffolds
220 of the GUY11 genome were identified among the 71 strains.

221 The PCA showed no genetic structure among individuals (Supp. Figure 2). Therefore, all strains
222 were kept in further analyses. LD decreased rapidly with physical distance between markers. The fit of
223 a logarithm regression $y=a.\ln(x) + b$ (with $a = -0.068$ and $b = 0.9434$) to the LD decay resulted in a half-
224 LD decay distance of 3.2 kb (Supp. Figure 3), showing a high recombination rate among the population.
225 Nucleotide diversity within the population, P_i (assessed on 31,834,555 nucleotides including 35,277
226 polymorphic sites) was estimated at 2×10^{-4} per site. This value agreed with the 2.11×10^{-4} value
227 estimated by Gladieux et al. (2018) for the *P. oryzae* recombinant lineage 1 to which the studied
228 population belongs.

229 *Phenotypic variability*

230 The strain effect on the production of microconidia was significant, as shown by analysis of
231 variance ($F=32.3$, $P<2\times 10^{-16}$, $Df=70$). Although the production of microconidia seemed to be more
232 variable between replicates for some strains than others (e.g. CH1019; Supp. Figure 4), the effect of the
233 technical replicate was not significant ($F=3.43$, $P=0.067$, $Df=1$). The batch effect, estimated using the
234 11 strains repeated in different batches, was significant ($F=63.3$, $P<2\times 10^{-16}$, $Df=8$). We corrected for this
235 batch effect using the lsmean method. The corrected value of microconidia production, hereafter
236 expressed in $\log(1 + \text{number of microconidia per mL})$, remained highly variable between strains (overall
237 mean, median, and variance: 12.2, 12.0, and 3.5 respectively; Figure 1). Some strains produced high
238 quantities of microconidia, e.g. CH1073, CH1152 and CH1110 (16.2, 16.3 and 17.6, respectively),
239 whereas others were low producers, e.g. CH1093 and CH1126 (8.3 and 8.8, respectively). The broad
240 sense heritability for the production of microconidia was estimated at 0.032.

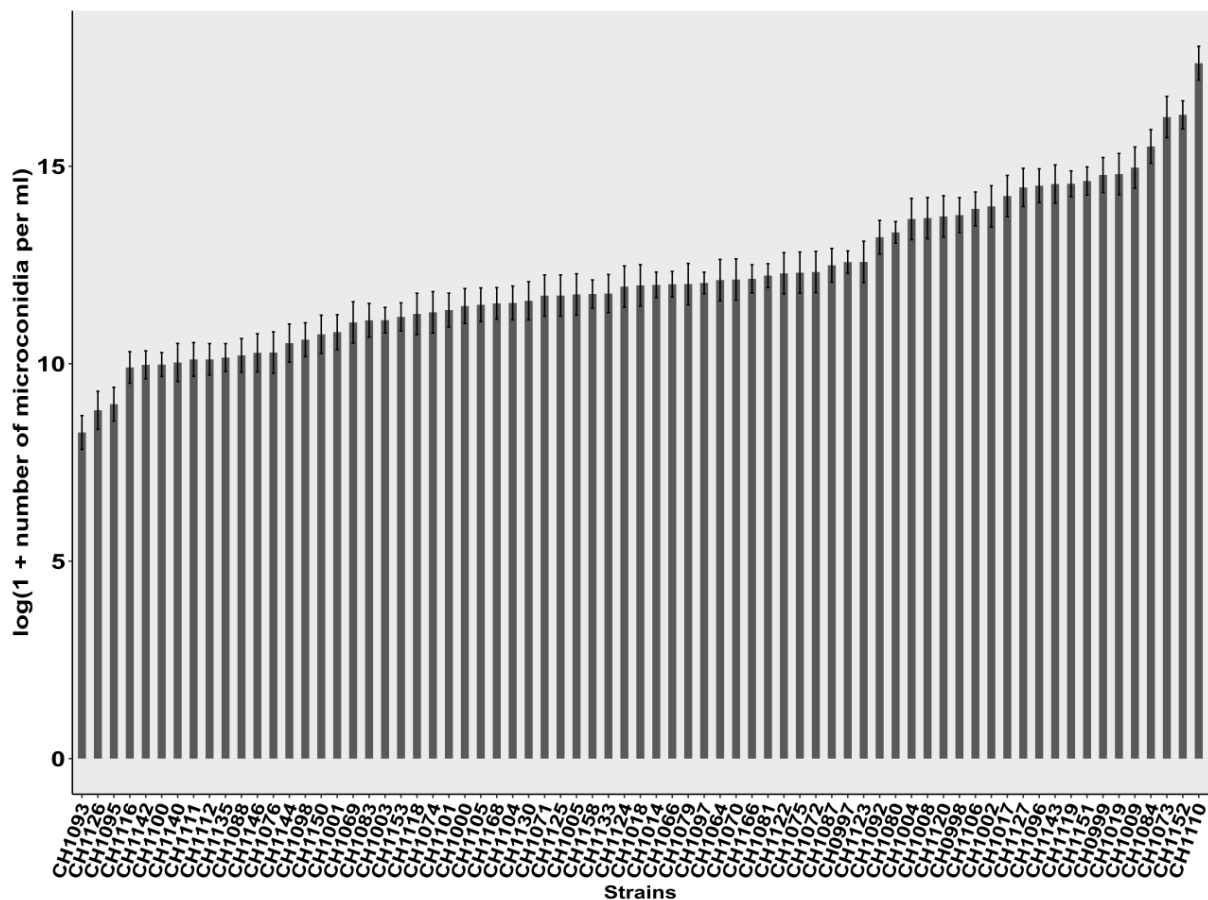


Figure 1: Production of microconidia of 71 strains of *P. oryzae*. The 71 strains were isolated from rice in Yule (Yunnan Province of China). The microconidia production is represented as $\log(1 + \text{number of microconidia per mL})$ corrected with the lsmean method. The standard deviation per strain is represented on each barplot.

241 *Association mapping*

242 GWAS analysis was performed on the 71 individuals using both a Mixed Linear Model (MLM)
243 and Multi-Locus Mixed model (MLMM) that included a kinship matrix. We used the mating-type
244 phenotype (verified biologically by *in vitro* crosses for all strains) as a positive control of GWAS. In the
245 70-15 reference genome (Dean et al., 2005) the MAT locus is located on chromosome 7 between
246 positions 772,852 and 778,251. The GWAs showed that in the GUY11 reference genome, the mating-
247 type phenotype was significantly associated with a peak on Scaffold 8 between positions 794,254 and
248 798,440. The alignment of this region against the 70-15 reference genome exactly matched with the
249 position of the MAT locus.

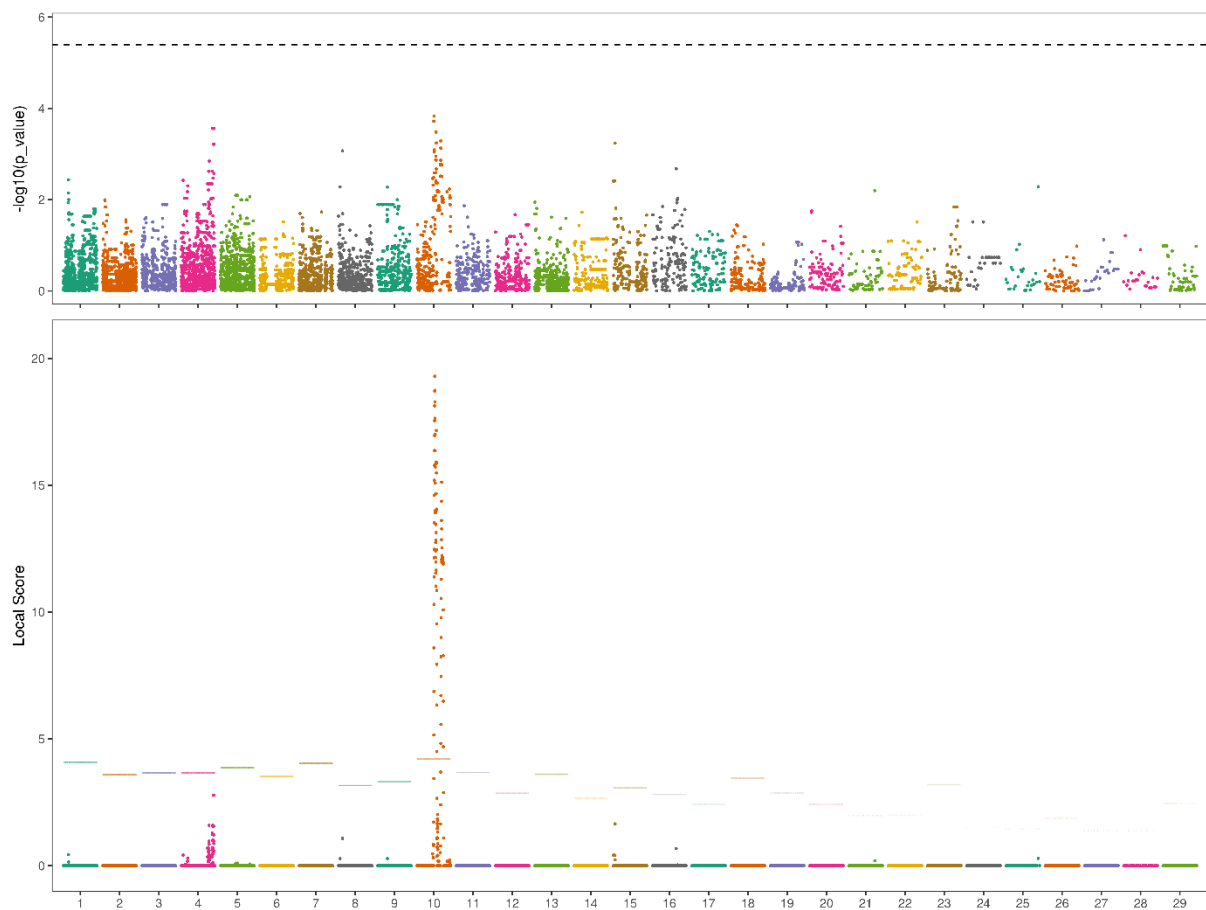


Figure 2: Genome-wide association mapping for microconidia production in *P. oryzae*. Upper panel: Manhattan plot of GWAS showing association p-values for each SNP marker along the 29 largest scaffolds covering 99% of GUY11 genome (one color per scaffold). P-values (vertical axis) are expressed in log. GWAS was performed with a Multi-Locus Mixed model including kinship matrix. The grey dashed line indicates the significance threshold after Dunn-Šidák correction with an α risk of 0.05 ($-\log[1-(1-0.05)^{1/14,000}] = 5.43$). Lower panel: Manhattan plot showing SNP marker local score values along the 29 largest scaffolds of GUY11. The local score values were calculated with $\xi = 2$. The horizontal dashed line corresponds to the scaffold-wide local score threshold with an α risk of 0.01.

250 We tested the association between the logarithm of the lsmean corrected values of production
251 of microconidia as phenotypic information [$\log(1 + \text{number of microconidia per mL})$], and the
252 polymorphism of 14,800 SNPs (biallelic, with no missing data) distributed on the 29 largest scaffolds
253 as genotypic information. GWAS with the MLM identified no SNP marker significantly associated to
254 the microconidia production when the Dunn-Šidák p-value threshold (here: $-\log[1-(1-0.05)^{1/14000}] = 5.43$)
255 was considered (Figure 2 upper panel). However, a region of sub-significant p-values was observed on
256 scaffold 10. Within this sub-significant region, the position with the highest $-\log_{10}(\text{p-value})$, equal to
257 3.83, was located at 927,900 bp. This region became significant when the local score approach was
258 applied. (Figure 2 lower panel), and was in fact composed of two very close significant peaks, from
259 position 924,790 to position 955,991 and from position 959,693 to position 967,996, respectively
260 (Figures 3). The non-significant sub-region separating the two peaks started at position 956,006 and
261 ended at position 959,601. In the first peak, the SNP marker with the highest local score (19.3) was
262 located at position 931,007. In the second peak, the highest value was 15.1 at the position 960,405. The
263 highest local scores were 4.6 and 3.6 greater than the local score threshold of 4.2, respectively. The two
264 peaks could be considered as one genomic region of 51.06 kb starting from position 917,291 and ending
265 at position 968,350. This region encompassed 166 SNP markers including 96 SNP markers significantly
266 associated to microconidia production (Figure 3). GWAS with MLM and subsequent local score method
267 led to same results as with MLM (Supp. Figure 5).

268 The clustering analysis of multilocus genotypes based on the 96 significant SNPs markers defined three
269 haplotypic groups (Figure 4A). The production of microconidia was significantly correlated with the
270 assignation of strains to these haplotypic groups (ANOVA: $F=5.35$, $P = 0.006$, $Df=2$). Haplotypic group
271 1 produced significantly more microconidia than haplotypic group 3 (Tukey test: $P = 0.005$), whereas
272 the production of microconidia by strains assigned to the haplotypic group 2 was not significantly

273 different from the two other groups (Tukey tests: 1 vs 2, $P = 0.44$; 2 vs 3, $P = 0.79$; Figure 4B). These
274 three haplotypic groups could be considered as three genotypes modulating microconidia production.

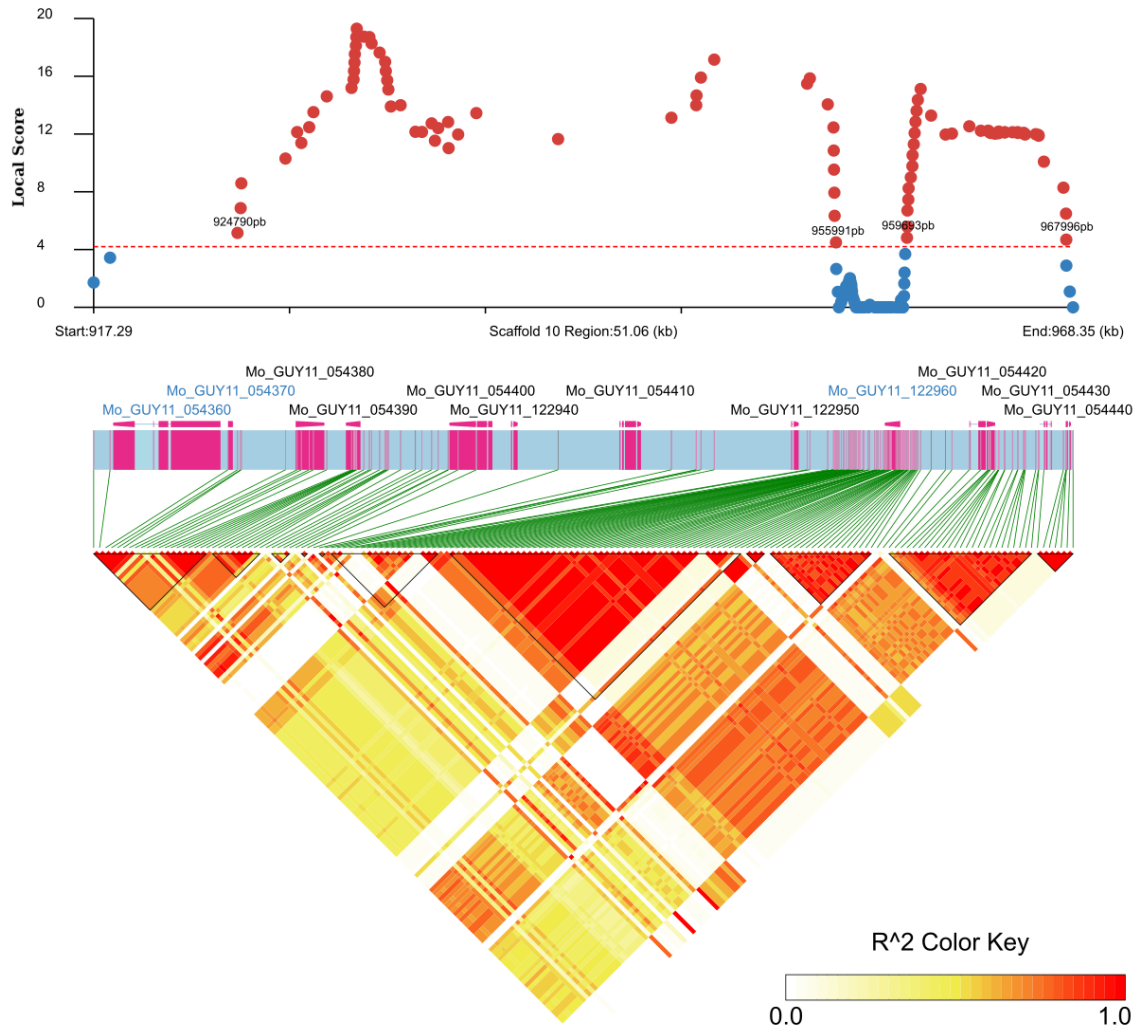


Figure 3: LD blocks contained in the associated region on scaffold 10 determined with the local score method. Upper panel: Manhattan plot of local score analysis on scaffold 10 (red dashed line: scaffold-wide local score threshold = 4.20). Significantly (respectively: non-significantly) associated SNP markers are indicated by red (respectively: blue) dots. Medium panel: Names and structural models of genes models located between the start (917.29 kb) and the end (968.35 kb) of the analyzed region on Scaffold 10. Names in black (respectively: blue) indicate genes located within the significant associated peaks (respectively: outside). Lower panel: LD blocks contained in the 51.06 kb region spanning 166 SNP markers.

275 *Candidate genes*

276 In GUY11, nine genes are predicted in the genomic region of scaffold 10 located between the
277 first and the last significant SNP markers (*ie* from 924,790 to 967,996; Figure 3). The corrected gene
278 models for each candidate gene in both reference genomes are summed up in Table 1. The markers with
279 the highest local score values were situated in the gene Mo_GUY11_054390 and close to the gene
280 Mo_GUY11_054420 for the first and second peaks respectively.

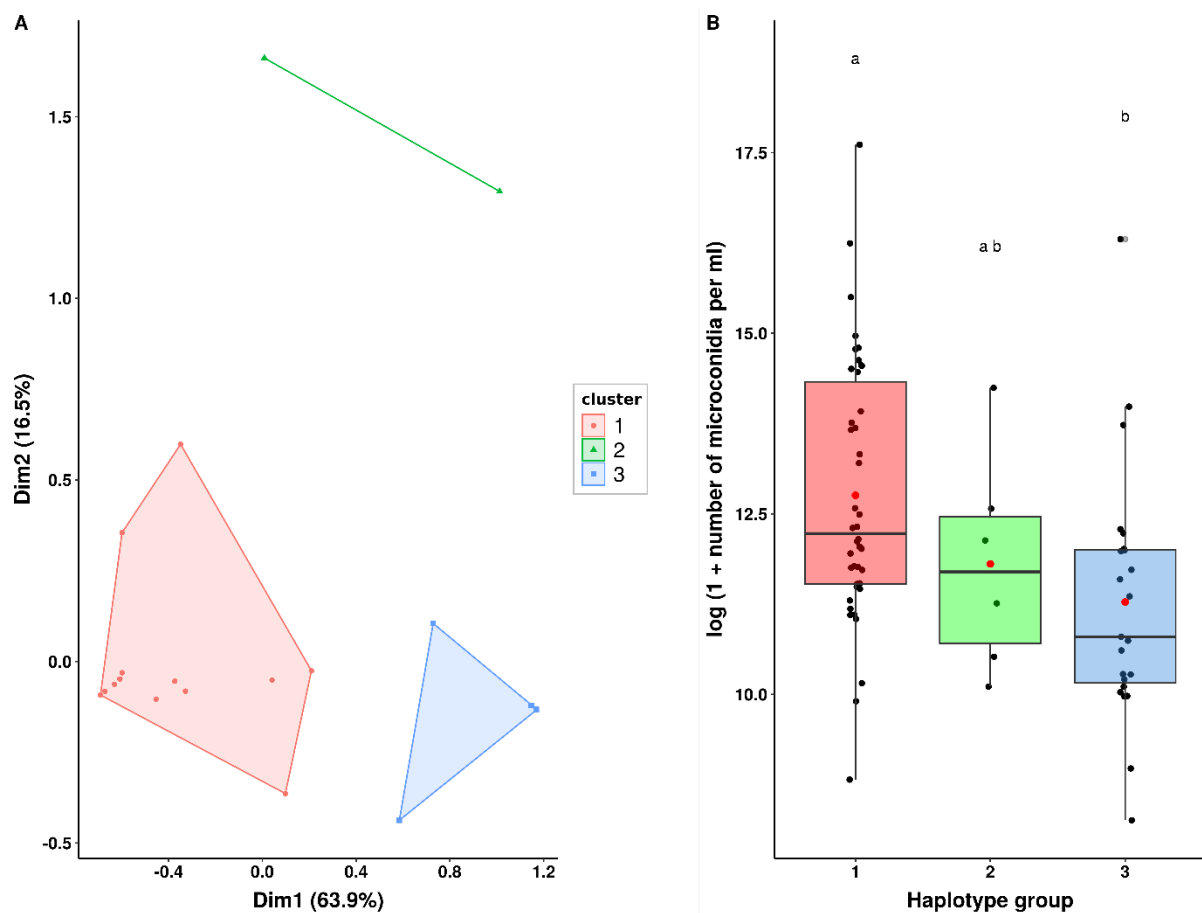


Figure 4: Clustering of strains based on multi-locus genotypes within the significantly associated region on scaffold 10. A: Multiple Correspondence Analysis performed on the 96 SNPs significantly associated to microconidia production, identifying three haplotypic groups (clusters 1, 2, 3). B: Boxplot of phenotypic values of strains assigned to each haplotypic group colored similarly to panel A (mean and median value for each group are shown by red point and horizontal black line, respectively).

281 **Table 1: Putative genes in the candidate region associated with microconidia production on GUY11 scaffold 10.**

GUY11 reference genome						70-15 reference genome						
Gene ID	Scaffold	Start	Stop	Length (bp)	Strand	Gene ID	Chromosome	Start	Stop	Length (bp)	Strand	Correction ¹
Mo_GUY11_054380	10	927822	929318	1497	-	MGG_02051	1	393105	395039	1935	+	No
Mo_GUY11_054390	10	930093	931202	1110	+	MGG_02050	1	391659	392768	1110	+	Yes
Mo_GUY11_054400	10	935840	938079	2240	+	MGG_02049	1	384782	387021	2240	+	No
Mo_GUY11_122940	10	939051	939395	345	+	MGG_16015	1	383466	383690	225	+	No
Mo_GUY11_054410	10	944594	946455	1862	-	MGG_02047	1	376406	378267	1862	+	Yes
Mo_GUY11_122950	10	953656	954046	391	-	MGG_13692	1	368872	369201	330	+	No
Mo_GUY11_122960*	10	958537	959340	804	+	MGG_16014	1	363621	364385	765	-	No
Mo_GUY11_054420	10	960874	964289	3416	-	MGG_02045	1	358633	362047	3415	+	Yes
Mo_GUY11_054430	10	966804	967511	708	+	MGG_16012	1	355412	356119	708	-	Yes
Mo_GUY11_054440	10	967968	968239	272	-	NA	NA	NA	NA	NA	NA	No

282 ¹: To overcome possible errors of annotation in GUY11 genome, gene models were, when necessary, corrected for start, stop and introns
 283 positions, so that the coding sequences (CDS) correspond to functional (non-interrupted) protein sequences.

284 *Gene located in a small sub-region where SNP markers were not significantly associated with the phenotype.

285 Among the nine predicted genes (whose sequences in the GUY11 genome are provided in
286 Supplementary text 2), only one had a predicted function in the 70-15 genome (Mo_GUY_054400 /
287 MGG_02049, encoding an Interferon-induced GTP-binding protein Mx2; Table 2). PFAM domains
288 were detected in five other genes: a protein Kinase domain in Mo_GUY_054380 (MGG_02051), a
289 GH43 Pc3Gal43A-like domain in Mo_GUY_054390 (MGG_02050), an oxidoreductase domain in
290 Mo_GUY11_122940 (MGG16015), a WH2 domain in Mo_GUY_054410 (MGG_02047), and a JmjC
291 domain in Mo_GUY_054420 (MGG_02045).

292 Only four genes showed polymorphism among the 71 strains in their CDS regions:
293 Mo_GUY11_054400, Mo_GUY11_054390, Mo_GUY11_054380, Mo_GUY11_054420 (Table 2).
294 Among these four genes, Mo_GUY11_054420, which contained the JmjC domain, was the most
295 polymorphic with 10 mutations. Mo_GUY_054420 was located 400 bp downstream of the SNP with
296 the highest local score of the second peak. The ratio of the number of mutations on the CDS length was
297 0.0030 whereas the mean of the nine genes was 0.0015. Mo_GUY11_054420 contained nine non-
298 synonymous mutations that lead to nine different protein sequences, whereas Mo_GUY_054380,
299 Mo_GUY11_054400, and Mo_GUY_054390 had 2, 1 and 4 non-synonymous mutations respectively
300 (Table 2). The non-synonymous mutation of Mo_GUY11_054400 was in the dynamin GTPase domain.

301 The region between the two peaks where SNP markers were not significantly associated with
302 the production of spermatia, contained one predicted gene (Mo_GUY11_122960). This gene, which
303 contains a DIOX_N domain, presented 11 mutations in the 71 strains with a ratio of 0.0137 mutation/bp.
304 Mo_GUY11_122960 contained seven non-synonymous mutations that lead to three different protein
305 sequences.

Table 2: Annotation of candidate genes

Gene ID	Synonymous / Non-Synonymous mutations	CDS length (pb)	Number of alternative proteic sequences	70-15 homologue	Protein function from 70-15 annotation	Predicted domain (Pfam / InterProScan / Uniprot search)	Putative function of homologue protein in other species	NCBI Domain search
Mo_GUY_054380	2/2	1497	3	MGG_02051	Uncharacterized	CAMK protein kinase	CAMK protein kinase	Serine-threo kinase, similar to protein CELE_K09C6 6
Mo_GUY_054390	4/4	1110	2	MGG_02050	Uncharacterized	GH43 Pc3Gal43A-like		
Mo_GUY_054400	0/1	2136	2	MGG_02049	Interferon-induced GTP-binding protein Mx2	Dynamin-type guanine nucleotide-binding domain, Interferon-induced GTP-binding protein Mx2	Putative vacuolar sorting protein VPS1, dynamin-2, Dynamin, GTPase domain, mitochondrial fission & membrane fusion	Proteins of the dynamin family catalyze membrane fission during clathrin-mediated endocytosis
Mo_GUY_122940	0/0	276	1	MGG_16015	Uncharacterized	Oxidoreductase	FAD-binding/transporter-associated domain-like	
Mo_GUY_054410	0/0	1737	1	MGG_02047	Uncharacterized	C2H2-type domain, WH2 domain	Actin-cytoskeleton organisation	
Mo_GUY_122950	0/0	306	1	MGG_13692	Uncharacterized		Breast carcinoma amplified sequence 2 (BCAS2)	
Mo_GUY_122960	4/7	804	3	MGG_16014	Uncharacterized	DIOX_N domain-containing protein		
Mo_GUY_054420	1/9	3315	9	MGG_02045	Uncharacterized	JmJ domain	JmjC domain-containing histone demethylation protein	The JmjC domain belongs to the Cupin superfamily.
Mo_GUY_054430	0/0	609	1	MGG_16012	Uncharacterized			Septal ring assembly protein ZapB; provisional
Mo_GUY_054440	0/0	177	1	NA	Uncharacterized	NA		

308 **Discussion**

309 In this study, we phenotyped 71 strains of a recombinant population of *P. oryzae* from Yule
310 (Yunnan Province, China) for male fertility, that is for the quantity of microconidia produced, and
311 showed that this character segregated in the population and was heritable. The use of GWAS combined
312 to local score approach was successful in detecting significant associations between this trait and one
313 genomic region. This region contains nine predicted genes, one of them being a candidate of particular
314 interest. Such a forward GWAS approach remains scarce in fungi, and to our knowledge, has never been
315 applied for studying fertility traits.

316 The population chosen for this study showed unambiguous genomic footprints of recombination
317 (as shown by LD decay analysis), which confirmed the results obtained by Thierry et al. (2022) with
318 GBS markers. The values of LD decay and nucleotide diversities were also in agreement with the
319 evaluations by Gladieux et al. (2018) for the rice-attacking lineage 1, to which the Yule population
320 belongs. The high recombination rate, high nucleotidic diversity and lack of population structure
321 observed in the Yule population confirmed that it is adequate for GWAS analysis.

322 The phenotyping of male gamete production showed that this trait, that we equate to male
323 fertility, segregated in the studied population and was heritable, albeit weakly. The significant batch
324 effect highlighted by the analysis of variance indicated that environmental variation could highly impact
325 the production of microconidia, at least in our experimental conditions. Previous studies on other fungal
326 species showed that the production of male gametes tightly relies on available resources (Debuchy et
327 al., 2010; Wilson et al., 2019). Resource availability could result in the complete lack of microconidia
328 production, and consequently, of sexual reproduction, in peculiar conditions. The potential role of
329 environmental conditions in the emergence and spread of clonal lineages in *P. oryzae* (Gladieux et al.,
330 2018; Saleh et al., 2014; Thierry et al., 2022) remains to be deciphered. The environmental effect also
331 likely contributes to the low heritability of male fertility observed in the studied population.

332 The low heritability of the studied trait might be one reason why classical GWAS failed to detect
333 any significant association between SNPs and the production of microconidia. In our study, the

334 combination of GWAS with the local score approach allowed to circumvent this limitation, proving that
335 local score is a powerful tool for the detection of the genomic bases of weakly heritable traits, and an
336 efficient alternative to QTL analyses. Indeed, QTL approaches are risky for traits related to fertility,
337 because the necessity to cross parents with extreme phenotypes (i.e. in our case: high- / low-
338 microconidia producers) jeopardizes the success of the cross itself and the obtention of enough progeny.
339 Furthermore, QTL approach provides a lower genomic resolution than GWAS and is restricted to the
340 allelic diversity of the two parents (Borevitz and Nordborg, 2003).

341 In *P. oryzae*, studies on microconidia production and more globally on male fertility are limited
342 to observation (Chuma et al., 2009) and demonstration of their fertilizing role (Lassagne et al., 2022).
343 In *P. oryzae*, mating type genes are required for the formation of perithecia (Wang et al., 2021). Genetic
344 bases of fertility have been scarcely explored for female fertility through perithecia production and asci
345 formation (Lee et al., 2021; Li et al., 2016). Furthermore, in these studies, the genetic bases involved in
346 fertility have been discovered by reverse genetic thanks to homologous genes already identified in other
347 fungal species (Peraza-Reyes and Malagnac, 2016). Here, combining GWAS with local score allowed
348 the discovery of one genomic. The genomic region significantly associated to microconidia production
349 highlighted by our GWAS analysis is a good candidate for further studies. This region, located on
350 GUY11 scaffold 10, is highly resolute (96 SNPs), and contains a limited number of coding-genes (9).

351 In the region of interest, Mo_GUY_054420 was the most promising candidate gene. This gene
352 and the marker most significantly associated to the phenotype were physically close (400bp) and
353 belonged to the same LD block. A more detailed analysis of multilocus genotypes in the LD block
354 showed that Mo_GUY_054420 was located in the region where markers are polymorphic between the
355 two genotypic clusters (cluster 1 and cluster 3) formed by strains that significantly differ for
356 microconidia production. In addition, Mo_GUY_054420 was the most polymorphic gene in the region
357 and showed also a high rate of non-synonymous mutations. Mo_GUY_054420 contains a JmjC domain.
358 Although the gene function remains unknown, the Jumonji family protein present in Eukaryotes has
359 been shown to be involved in chromatin regulation and in many signaling pathways (Takeuchi et al.,
360 2006). In *P. oryzae*, the gene MoJMJ1 (MGG_04878) encodes a histone demethylase containing a JmjC

361 domain. Deletion of MoJMJ1 reduced mycelial growth and asexual spore production, altered germ-tube
362 formation and suppressed appressorium formation (Huh et al., 2017). In the region of interest, we also
363 found the Mo_GUY_054380 gene whose role deserves to be functionally explored with regards to
364 spermatia production. This gene encodes a putative Ca²⁺/calmodulin-dependent protein kinase
365 (CAMK), a class of serine/threonine-specific protein kinases that phosphorylates transcription factors.
366 Some of the CAMK responding genes were shown to regulate cell life cycle and cytoskeleton network
367 (Berchtold and Villalobo, 2014). These authors also identified the gene Mo_GUY11_054400 which
368 encodes a putative dynamin GTPase potentially related to cytoskeleton organization. Proteins endowed
369 with a dynamin domain are essential to membrane fusion and fission, from endocytosis to organelle
370 division. They are involved in microtubules organization and clathrin-mediated cell membrane
371 invaginations to form budding vesicles (Antonny et al., 2016), a process that may be at work during
372 formation of spermatia from syncytial hyphae. Importantly, none of the five genes of the interval of
373 interest showing polymorphism in their CDS were previously identified as involved in male gametes
374 differentiation.

375 This study confirmed that the Yule population, and more generally, population from the
376 recombinant rice-attacking lineage 1, are male and female fertile (Saleh et al., 2014; Gladieux et al.,
377 2018; Thierry et al., 2022). Previous studies also showed that the three other rice-attacking lineages
378 found worldwide were clonal, and that populations from these lineages exhibited low to null levels of
379 female fertility and a single mating type. Thus, these populations were considered to reproduce only
380 asexually. It could be interesting to phenotype populations from these clonal lineages for male fertility
381 to test whether they have completely lost the biological ability to produce male gametes, and therefore,
382 to perform sexual reproduction. In parallel, comparing the polymorphism in the candidate genes
383 controlling male fertility in recombinant populations from lineage 1 and in clonal populations from
384 lineages 2-4 would contribute to a better understanding of the causes, stochastic or adaptive, of the loss
385 of fertility that accompanied the migration of clonal lineages all around the world.

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391

392 **Data accessibility statement**

393 Raw reads will be made accessible at the European Nucleotide Archive (accession no. PRJEB67445).
394 Scripts used for mapping, SNP calling, population genomics analyses and statistical analyses of
395 phenotypic data, are provided as supplementary material.

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