ECOLOGICAL DIFFERENTIATION AMONG GLOBALLY DISTRIBUTED LINEAGES OF THE RICE BLAST

FUNGUS *PYRICULARIA ORYZAE*

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ABSTRACT

Many invasive fungal species coexist as multiple lineages on the same host, but the factors underlying

the origin and maintenance of population structure remain largely unknown. Here, we analyzed

genetic and phenotypic diversity in isolates of the rice blast fungus (*Pyricularia oryzae*) covering a

broad geographical range. We showed that the four lineages of *P. oryzae* were found in areas with

different prevailing environmental conditions and types of rice grown, indicating niche separation.

Pathogenicity tests revealed that specialization to rice subspecies contributed to niche separation

between lineages, and differences in repertoires of putative pathogenicity effectors were consistent

with differences in host range. Experimental crosses revealed that female sterility and early post-

mating genetic incompatibilities acted as strong barriers to gene flow between these lineages. Our

results demonstrate that the spread of a pathogen across heterogeneous habitats and divergent

populations of a crop species can lead to niche separation and reproductive isolation between

distinct invasive lineages.

INTRODUCTION

Invasive fungal plant pathogens constitute a major threat to ecosystem health and agricultural production (Desprez-Loustau *et al.* 2007; Rizzo 2005; Strange & Scott 2005). Three decades of molecular plant pathology have revealed a seemingly inexhaustible stream of disease outbreaks or significant changes in the geographic location of diseases and host ranges of pathogens. In addition to the burden of emerging fungal infections, most major crops or domesticated trees are colonized by pathogens that are so widespread and have been present for so long, that they do not necessarily come to mind when one thinks about invasive fungi (Palm 2001; Yarwood 1970). If fungal pathogens are to spread successfully over their host species distribution, they must overcome a series of barriers to invasion related to dispersal, host availability, competition with other microbes, and abiotic conditions (Gladieux *et al.* 2015). An understanding of the factors underlying the success of invasive fungal pathogens is essential to protect world agriculture against future pandemics (Thrall *et al.* 2011).

Invasive fungal plant pathogens display a range of population structures, reflecting their highly diverse life-history strategies and invasion histories (Barrett *et al.* 2008; Gladieux *et al.* 2015; Hessenauer *et al.* 2020). Demographic events (e.g. population bottlenecks, secondary contact) and natural selection during spread across heterogeneous environments may be associated with intricate changes in demography and life history traits, or may cause such changes. Shifting to a new host is a primary life history trait change in invasive fungal pathogens, and a major route for the emergence of disease. Changes in host range are facilitated by certain features, such as mating within or on their hosts and strong selective pressure on a limited number of genes (Giraud *et al.* 2010; Gladieux *et al.* 2011a). Reproduction within, or on plant hosts favors assortative mating with respect to host use, leading to a strong association between adaptation to the host and reproductive isolation (Giraud *et al.* 2006; Servedio *et al.* 2011) and the differentiation of pathogen populations adapted to different host populations, varieties or species (Lemaire *et al.* 2016; Mercier *et al.* 2021; Peever 2007;

Stukenbrock *et al.* 2011). Such changes in pathogen host range can result from immune-escape

mutations at a small number of genes encoding small secreted proteins called effectors, which

enable microbes to influence the outcome of host-pathogen interactions for their own benefit, and

which may be subject to surveillance by the plant immune system (Couch *et al.* 2005; Schulze-Lefert

& Panstruga 2011a; Uhse & Djamei 2018). Another frequent change in the life history traits of

invasive fungal pathogens is the loss of sexual reproduction (Taylor *et al.* 2015; Taylor *et al.* 1999).

Invasive fungal pathogens may develop a clonal population structure over their entire introduced

range or over parts of that range ((Gladieux *et al.* 2015) and references therein). The immediate

causes of a the loss of sexual reproduction include a lack of compatible mating partners in the

introduced range (i.e. a lack of compatible mating types) (Hessenauer *et al.* 2020), hybridization and

hybrid sterility (Li *et al.* 2019; Lin & Heitman 2006), and a lack of compatible alternate hosts for

sexual reproduction (Fellers *et al.* 2021). Mating type loss may be driven by the extreme population

bottlenecks associated with the colonization of a new host or new area. Shifts to asexual

reproduction may also result from selection against reproduction (Gladieux *et al.* 2015), particularly

for pathogens of domesticated crops and trees, due to the availability of large, homogeneous host

populations, releasing constraints that maintain sexual reproduction in the short term (de Vienne *et*

al. 2013). These conditions may favor the asexual spread of new lineages carrying adaptive allelic

combinations requiring shelter from recombination (Salcedo *et al.* 2017). Finally, changes in life

history traits can also result from variations in environmental conditions, including temperature in

particular, during the course of invasive spread (Frenkel *et al.* 2010; Mboup *et al.* 2012). Deciphering

the complex interplay between changes in population structure and life history strategies that

accompanies the spread of invasive fungi requires deep sampling the and integration of genetic,

phenotypic, and environmental information throughout the native and invasive ranges.

Pyricularia oryzae (Ascomycota; syn. *Magnaporthe oryzae*) is a widespread model plant

pathogen displaying population subdivision. *Pyricularia oryzae* is best known as the pathogen causing

rice blast, which remains a major disease of Asian rice (*Oryza sativa*), but it is also a major constraint

on the production of wheat (*Triticum*), finger millet (*Eleusine coracana*) and Italian millet (*Setaria*

italica), a significant disease of ryegrass and an emerging pathogen of maize (*Zea maydis*) (Farman *et*

al. 2017; Inoue *et al.* 2017; Islam *et al.* 2016; Milazzo *et al.* 2019; Pordel *et al.* 2021; Tanaka *et al.*

2009). An in-depth genomic analysis of a large worldwide collection of *P. oryzae* isolates from 12 host

plants revealed the existence of 10 distinct *P. oryzae* lineages, each associated mostly with a single

main cereal crop or grass host (Gladieux *et al.* 2018a). Sequence divergence between lineages was

only slight, of the order of about 1% (Gladieux *et al.* 2018a), and analyses of gene flow and admixture

provided evidence that the different host-adapted lineages were connected by relatively recent

genetic exchanges, and therefore corresponded to a single phylogenomic species (Langner *et al.*

2018; Valent *et al.* 2019). The reproductive biology of *P. oryzae* is typical of plant pathogenic

Ascomycetes. *Pyricularia oryzae* has a haplontic life cycle. Thus, the multicellular state is haploid, and

the breeding system is heterothallic, with mating occurring only between haploid individuals of

opposite mating types. *Pyricularia oryzae* is hermaphrodite, but successful mating requires that at

least one of the partners to be capable of producing female structures (i.e. "female-fertile").

The lineage of *P. oryzae* that infects rice is the most widely studied at the population level,

due to its agricultural importance. Analyses of populations from commercial fields and trap nurseries

based on molecular markers and mating assays have consistently shown that non-Asian populations

of the rice-infecting lineage harbor a single mating-type and are clonal, with sexual reproduction

never observed in the field (Tharreau *et al.* 2009; Zeigler 1998). Populations with female-fertile

strains, a coexistence of both mating types and signatures of recombination have been observed

exclusively in Asia (Kumar *et al.* 1999; Saleh *et al.* 2014; Saleh *et al.* 2012b; Tharreau *et al.* 2009;

Zeigler 1998). More recent population genomic studies of the rice-infecting lineage have revealed

genetic subdivision into four main lineages, estimated to have diverged approximately 1000 years ago

and displaying contrasting modes of reproduction (Gladieux *et al.* 2018b; Latorre *et al.* 2020; Zhong *et*

al. 2018). Only one of the four lineages that infects rice (lineage 1, prevailing in East and Southeast

Asia) displays a genome-wide signal of recombination, a balanced ratio of mating type alleles and a

high frequency of fertile females, consistent with sexual reproduction (Gladieux *et al.* 2018b; Saleh *et*

al. 2014). All the other lineages (lineages 2, 3 and 4, present on several continents) have clonal

population structures, highly biased frequencies of mating types and only a few fertile females,

suggesting that reproduction in these lineages is strictly asexual (Gladieux *et al.* 2018b; Saleh *et al.*

2014). The recombining lineage has a nucleotide diversity four times greater than that of the clonal

lineages (Gladieux *et al.* 2018c). In theory, gene flow remains possible between the majority of the

lineages, because there are rare fertile strains and some lineages have fixed opposite mating types.

However, while resequencing studies have revealed signatures of recent shared ancestry between

clonal lineages and the recombinant lineage from South-West Asia, this is not the case between the

different clonal lineages despite their wide and overlapping distributions, suggesting a lack of recent

gene flow between lineages of opposite mating types (Gladieux *et al.* 2018b). The loss of female

fertility is probably a key factor in the emergence of clonal lineages, inducing a shift towards strictly

asexual reproduction, and protecting populations adapting to new conditions from maladaptive gene

flow from ancestral, recombining populations (Giraud *et al.* 2010). This raises questions about the

nature of the factors underlying the emergence of different rice-infecting lineages in *P. oryzae*, and

contributing to their maintenance, because, according to the competitive exclusion principle,

different lineages should not be able to co-exist on the same host (Abbate *et al.* 2018; Giraud *et al.*

2017; Hardin 1960). Previous studies have helped to cement the status of *P. oryzae* as a model for

studies of the population biology of fungal pathogens, but most efforts to understand the population

structure of the pathogen have been unable to provide a large-scale overview of the distribution of

rice-infecting lineages and of the underlying phenotypic differences, because the number of genetic

markers was limited (Saleh *et al.* 2014; Tharreau *et al.* 2009), the number of isolates was relatively

small (Gladieux *et al.* 2018b; Zhong *et al.* 2018) and because few phenotypic traits were scored (Saleh

et al. 2014; Tharreau *et al.* 2009; Zhong *et al.* 2018).

We provide here a detailed genomic and phenotypic overview of natural variation in rice-

infecting *P. oryzae* isolates sampled across all rice-growing areas of the world. With a view to

inferring and understanding the global population structure of rice-infecting *P. oryzae*, we analyzed

genetic variation, using genotyping data with a high resolution in terms of genomic markers and

geographic -coverage, and measured phenotypic variation for a representative subset of the sample

set. We also characterized the genetic variability and content of repertoires of effectors –secreted

proteins that modulate host responses – using whole-genome resequencing data for a representative

set of isolates. We addressed the following specific questions: (i) How many different lineages of *P.*

oryzae infect rice? (ii) Do lineages of *P. oryzae* display footprints of recombination? (iii) Do lineages of

P. oryzae present evidence of recent admixture? (iv) Are the different lineages sexually competent

and compatible? (v) Are the different lineages differentially adapted to host and temperature? (vi)

Can we find differences in the number, identity and molecular evolution of putative effectors

between lineages? (vii) Can we find differences in the geographic and climatic distribution of

lineages?

RESULTS

The rice blast pathogen comprises one recombining, admixed lineage and three clonal, non-

admixed lineages

We characterized the global genetic structure of the rice blast pathogen, using an Illumina

genotyping beadchip to score 5,657 SNPs distributed throughout the genome of 886 *P. oryzae*

isolates collected from cultivated Asian rice in 51 countries (Supplementary file 1). The final dataset

included 3,686 SNPs after the removal of positions with missing data, which identify 264 distinct

multilocus genotypes out of 886 isolates. The genotyping error rate was below 0.62%

(Supplementary file 1). Clustering analyses based on sparse nonnegative matrix factorization

algorithms, as implemented in the sNMF program (Frichot et al., 2014), revealed four clusters,

hereafter referred to as "lineages" (Figure 1C). The model with K=4 clusters was identified as the best

model on the basis of the cross-entropy criterion, as models with K>4 induced only a small decrease

in cross-entropy, suggesting that K=4 captured the deepest subdivisions in the dataset (Figure 1 –

figure supplement 1). The neighbor phylogenetic network inferred with SPLITSTREE also supported the

subdivision into four widely distributed lineages, with long branches separating three peripheral

lineages branching off within a central lineage (Figure 1A; 1B). Principal Component Analysis (PCA)

revealed four groups coinciding with the four lineages identified with phylogenetic network and

clustering analyses. A comparison with previous findings revealed that the central lineage in the

Neighbor-net network and PCA corresponded to the combination of recombining lineage 1 and

lineages 5 and 6, represented by two and one individuals, respectively, in a previous study (Gladieux

et al. 2018b) (Supplementary file 2). This central lineage is referred to hereafter as lineage 1. The

three peripheral lineages in the Neighbor-net network and PCA corresponded to the previously

described lineages 2, 3 and 4 (Gladieux *et al.* 2018b) with lineages 2 and 3 similar to lineages B and C,

respectively, identified on the basis of microsatellite data in a previous study (Saleh *et al.*

2014)(Supplementary file 2). Genetic differentiation between the four lineages was strong and

significant (Weir and Cockerham's F_{ST}> 0.54), indicating strong barriers to gene flow between the

lineages. All genotypes from lineages 2 to 4 had high proportions of membership q in a single cluster

($q > 0.89$), whereas shared ancestry with the other three lineages was detected in lineage 1, with

32% of genotypes having $q > 0.10$ in lineages 2-4 (Figure 1C). Admixture may account for the lower

F_{ST} observed in comparisons between lineage 1 and the other lineages.

We detected no footprints of recombination in lineages 2, 3 and 4 with three different tests

(pairwise homoplasy index (Bruen *et al.* 2006), maximum χ^2 (Maynard Smith 1992) and neighbour

similarity score (Jakobsen & Easteal 1996)) and a null hypothesis of clonality (Table 1), confirming

previous findings obtained with smaller datasets (Gladieux *et al.* 2018b; Latorre *et al.* 2020; Saleh *et*

al. 2014; Zhong *et al.* 2018). The alignment of genotypes from lineages 2, 3 and 4 in the PCA (Figure

1C), their lack of linkage disequilibrium decay with increasing physical distance (Supplementary file

3), and their low proportions of homoplastic mutations (Table 1), provided additional evidence for an

absence of recombination in these lineages, contrasting with lineage 1.

Using whole-genome data for isolates from lineages of *P. oryzae* infecting other grasses and

cereals, we showed that our set of Infinium SNPs could be used to identify isolates not belonging to

the rice-infecting lineage, although it was unable to differentiate between the different host-specific

lineages of *P. oryzae* (Appendix 1). Clustering analysis identified to two isolates from West Africa

(BN0019 and BF0072, clonal group 18 [Supplementary file 1]) corresponding to a lineage different

from the rice lineage and possibly resulting from ancient admixture between other lineages

(Appendix 1). However, on inclusion of these divergent isolates in our dataset, we detected no fixed

differences with respect to the rest of the isolates from lineage 1, and these divergent isolates did

not form a separate group in clustering analyses. They are not, therefore, likely to have affected on

our main conclusions on population subdivision and mode of reproduction.

Figure 1. Rice-infecting *P. oryzae* populations are divided into four major lineages. Population subdivision was inferred from 264 distinct *P. oryzae* genotypes, representing 886 isolates, and the four lineages were represented in different colors. **(A)** Neighbor-net phylogenetic network estimated with SPLITSTREE; reticulations indicate phylogenetic conflicts caused by homoplasy. **(B)** Geographic distribution of the four lineages identified with SPLITSTREE (A) and sNMF (D), with disk area proportional to sample size. **(C)** Principal Component Analysis with colors indicating lineages identified with SPLITSTREE (A) and sNMF (D). **(D)** Ancestry proportions in *K*=4 clusters, as estimated with sNMF software; each multilocus genotype is represented by a vertical bar divided into four segments indicating membership in *K*=4 clusters.

Figure 1 – figure supplement 1. Cross-entropy (CE) as a function of the number of clusters K modeled in sNMF analyses of population subdivision.

Figure 1 – source data 1.

Multilocus genotypes for neighbor-net phylogenetic inference, principal component analysis and recombination analyses.

Figure 1 – source data 2.

Ancestry proportions in K=4 sNMF clusters.

Geographic structure and admixture within recombining lineage 1

Lineage 1 was the only lineage for which there was population genetic evidence of recombination in the PHI test (Table 1). Most of the lineage 1 isolates were collected in Asia (79%), but the lineage was present on all continents in which the pathogen was sampled (Europe: one isolate; North America:

10, Central and South America: 7, Africa: 22) (Figure 2). Clustering analysis on this single lineage with

sNMF detected four clusters with different geographic distributions (Figure 2). Estimates of

differentiation were lower between the clusters within lineage 1 (F_{ST} < 0.49) than between the main

lineages (F_{ST} > 0.54), consistent with a longer history of restricted gene flow between the main

lineages and a more recent history of admixture between clusters within lineage 1 (Supplementary

file 3; Figure 1C).

Two clusters within lineage 1 (referred to hereafter as *Baoshan* and *Yule*) consisted mostly of

isolates sampled from two different sites in the Yunnan province of China (Baoshan and Yule,

respectively, located about 370 km apart). The third cluster consisted mostly of isolates from Laos

and South China (referred to hereafter as *Laos*), and the fourth brought together 95% of the isolates

from lineage 1 collected outside Asia (referred to hereafter as *International*). In Asia, the *International*

cluster was found mostly in the Yunnan province of China, India and Nepal (Figure 2). PHI tests for

recombination rejected the null hypothesis of clonality for all four clusters (Table 1). Admixture was

widespread in lineage 1 (Figure 2), with most of the isolates (78%) displaying membership

proportions with *q* > 0.10 for two or more clusters (Figure 2). Only five genotypes were detected in

multiple countries (genotype ID [number of countries]: 2 [6], 18 [2], 58 [2], 98 [3] and 254 [3],

corresponding to 41 isolates in total). All these genotypes belonged to the *International* cluster.

Figure 2. Population subdivision in lineage 1. **(A)** Neighbor-net phylogenetic network estimated with SPLITSTREE; reticulations

indicate phylogenetic conflicts caused by homoplasy. **(B)** Principal Component Analysis with colors indicating clusters

identified with SPLITSTREE (A) and sNMF (D). **(C)** Geographic distribution of the four clusters identified with sNMF (D), with

disk area proportional to number of isolates. **(D)** Ancestry proportions in four clusters, as estimated with sNMF; each

multilocus genotype is represented by a vertical bar divided into four segments, indicating membership in K=4 clusters. **(E)**

Number of isolates and their geographic origin for each multilocus genotype of lineage 1. Panels D and E share x-axis (each

vertical bar represents a different multilocus genotype).

Figure 2 – source data 1.

Multilocus genotypes for neighbor-net phylogenetic inference.

Figure 2 – source data 2.

Ancestry proportions in K=4 sNMF clusters.

Figure 2 – source data 3.

Geographic distribution of clones (i.e. multilocus genotypes repeated multiple times).

Table 1. Tests for recombination (null hypothesis: clonality), distribution of mating types, fertile females, and proportion of

homoplastic polymorphisms. Analyses were conducted for four lineages, and for clusters within lineage 1.

Lineage/Cluster	PHI (p-value)	Max χ^2 (p-value)	NSS (p-value)	Homoplastic sites (%)	Mat1.1/Mat1.2	Female-fertile isolates (%)
1	<0.0001	<0.0001	<0.0001	1388/2214 (63)	53/47	43
Baoshan	<0.0001	<0.0001	<0.0001	NC	69/31	67
International	<0.0001	<0.0001	<0.0001	NC	55/45	4
Laos	<0.0001	<0.0001	<0.0001	NC	68/32	37
Yule	<0.0001	<0.0001	<0.0001	NC	40/60	79
2	0.25	0.32	0.72	1/391 (0)	97/3	0
3	0.18	0.35	0.88	0/408 (0)	3/97	7
4	0.22	0.30	0.51	0 (218)	97/3	0

NC, not computable because clusters within lineage 1 are polyphyletic in the phylogeny used to estimate the proportion of

homoplastic sites. Tests for recombination: Pairwise homoplasy index (PHI; (Bruen *et al.* 2006)), Maximum χ^2 (Max χ^2,

(Maynard Smith 1992)) and Neighbour Similarity Score (NSS; (Jakobsen & Easteal 1996)). Source datafiles: Figure 1 – source

data 1 for PHI, Max χ^2, NSS; Table 1 – source data 1, Table 1 – source data 2 and Table 1 – source data 3 for the proportion

of homoplastic sites. Information about mating types and female fertility is available in Supplementary file 1.

Table 1 – source data 1.

RAxML genealogy for homoplasy analysis, estimated from Figure 1 – source data 1.

Table 1 – source data 2

Pairs of nodes included in homoplasy analysis, as determined from Table 1 – source data 1.

Table 1 – source data 3

Result of 'Trace all characters' analysis conducted with MESQUITE.

Reproductive barriers caused by female sterility and postmating genetic incompatibilities

Our analysis of genotyping data revealed the existence of a recombinant population from which

several clonal lineages have more recently emerged. This raised the question as to why these

populations became clonal, with no further gene flow. We addressed this question by determining

the capacity of the different lineages to engage in sexual reproduction, by characterizing the

distribution of mating types and the ability to produce female structures (i.e. female fertility). Using

in vitro crosses with tester strains, we found that lineages 2, 3 and 4 were composed almost

exclusively of a single mating type (97% of lineage 2 and 4 isolates tested carried the *Mat1.1* allele;

97% of lineage 3 isolates carried the *Mat1.2* allele), and had only small proportions (0-7%) of isolates

with female fertility (Table 1). By contrast to these clonal lineages, the mating type ratio was more

balanced in lineage 1 (52% of *Mat1.1*; Table 1), with *Mat1.1*/*Mat1.2* ratios ranging from 40/60 in the

Yule cluster to 69/31 in the *Baoshan* cluster. Female fertility rates were higher in three of the clusters

in lineage 1 (*Yule*: 79%; *Baoshan*: 67%; *Laos*: 37%; Table 1), but that in the International cluster (4%

female fertility) was as low as those in the clonal clusters (Table 1). These observations indicate that

there is a potential for sexual reproduction between lineages with compatible mating types and,

albeit rare, fertile females, but not between lineages 2 and 4, due to the low female fertility rates

and highly biased mating type ratios.

(A) Production of perithecia

(B) Production of germinating ascospores

Figure 3. Success of crosses between lineages 2-4 and clusters within lineage 1 with (A) proportion of crosses producing at least one perithecium, (B) scoring of ascus formation and ascospore germination for a subset of crosses.

Figure 3 – Source data 1

Production of perithecia

Figure 3 – Source data 2

Production of germinating ascospores

We further assessed the likelihood of sexual reproduction within and between lineages, by evaluating the formation of sexual structures (perithecia), the production of asci (i.e. meiotic octads within the perithecia) and the germination of ascospores (i.e. meiospores in meiotic octads) in

crosses between randomly selected isolates of opposite mating types from all four lineages (Figure 3).

This experiment revealed a marked heterogeneity in the rate of perithecium formation across

lineages. Three clusters in lineage 1 had the highest rates of perithecia formation, with isolates in the

Yule cluster, in particular, forming perithecia in 93% of crosses with isolates from the same cluster,

and in more than 46% of crosses with isolates from other lineages (Figure 3A). Isolates from lineages

2, 3 and 4 could only be crossed with isolates from other lineages, given their highly biased mating

type ratios, and the percentage of these crosses producing perithecia was highly variable (ranging

from 0% to 83%, depending on the lineages involved; Figure 3A). The rate of perithecium formation

was similar in crosses involving the *International* cluster of lineage 1 and in crosses involving clonal

lineages 2-4, and none of the crosses attempted between isolates from the *International* cluster led

to perithecium formation (Figure 3A). Perithecium dissection for a subset of crosses involving some of

the most fertile isolates revealed that most inter-lineage crosses produced perithecia that did not

contain asci or contained asci with low ascospore germination rates (Figure 3B). Whereas 100% of

crosses between isolates of the *Yule* cluster produced numerous germinating ascospores, ascospore

germination rates were only 33%, 56% and 7% in *Yule* x lineage 2, *Yule* x lineage 3 and *Yule* x lineage 4

crosses, respectively. Together, these results indicate that the clonal structure of lineages 2-4 is

caused by highly biased mating types and female sterility, and that the three clonal lineages and the

International cluster of lineage 1 are isolated from each other by strong pre- and early-postmating

barriers, including breeding system isolation (differences in mating type and female sterility), and a

combination of gametic incompatibility, zygotic mortality or hybrid non-viability. However, three of

the clusters in lineage 1 had biological features consistent with an ability to reproduce sexually,

suggesting possible sexual compatibility with clonal lineages 2-4 and the *International* cluster.

Specialization to temperature conditions and rice subspecies

Given the very broad distribution of *P. oryzae* and the strong environmental heterogeneity with

which it is confronted in terms of the nature of its hosts and the climate in which it thrives, we

evaluated the variation of pathogen fitness over different types of rice and temperature conditions.

We measured growth rate and sporulation rate of representative isolates cultured at different

temperatures to test the hypothesis of adaptation to temperature. Experiments were conducted

with 41 isolates (lineage 1 [yule cluster]: 11; lineage 2: 10; lineage3: 10; lineage 4: 10). For all

lineages, mycelial growth rate increased with incubation temperature, although this trend was more

visible from 10°C to 15°C (increased mean mycelial growth of +2.22 mm/day) than from 25°C to 30°C

(+0.05 mm/day) (growth curves and a full statistical treatment of the data are presented in

Supplementary file 5; data are reported in Supplementary file 6). Fitting a linear mixed-effects model

with incubation time, experimental replicate, and lineage of origin as fixed effects and isolate as a

random effect, revealed a significant lineage effect at each incubation temperature [10°C:

$F_{(1,364)}=7988$, $p<0.001$; 15°C: $F_{(1,419)}=33161$, $p<0.001$; 20°C: $F_{(1,542)}=40335$, $p<0.001$; 25°C:

$F_{(1,413)}=30156$, $p<0.001$; 30°C: $F_{(1,870)}=52681$, $p<0.001$]. Comparing least-squares means resulting

from linear mixed-effects models at each temperature revealed a significantly lower growth rate of

lineage 4 at 10°C relative to other lineages and a significantly higher growth rate of lineage 1 at 15°C,

20°C, 25°C and 30°C relative to other lineages. Almost no sporulation was detected at 10°C after 20

days of culture, with the few spores observed being not completely formed and divided by only one

septum instead of two septa in mature conidia (sporulation curves and a full statistical treatment of

the data are presented in Supplementary file 6; data are reported in Supplementary file 8). For all

lineages, sporulation rates increased with temperature from 15 to 25 °C and dropped at 30°C,

although isolates were cultured for only 7 days only at this temperature, versus 10 days at other

temperatures, because by this stage, the mycelium had already reached the edges of the plates.

Significant lineage effects were observed at 10°C, 15°C and 30°C (Kruskal-Wallis tests; 10°C:

$H(3)=9.61$, $p=0.022$; 15°C: $H(3)=16.8$, $p<0.001$; 30°C: $H(3)=8.96$, $p=0.030$). Pairwise non-parametric

comparisons of lineages based on Dunn's test revealed significant differences between lineages 1

and 2 at 10°C, between lineage 1 and lineages 3 and 4 at 15°C, and between lineage 1 and 4 at 30°C.

Together, measurements of mycelial growth and sporulation at different temperatures revealed

differences in performance between lineages, but no clear pattern of temperature specialization.

We assessed the importance of adaptation to the host, by challenging 45 rice varieties

representative of the five main genetic groups of Asian rice (i.e. the temperate japonica, tropical

japonica, aus, aromatic and indica subgroups of *Oryza sativa*) with 70 isolates representative of the

four lineages of *P. oryzae* and the four clusters within lineage 1 (Supplementary file 9). Interaction

phenotypes were assessed qualitatively by scoring resistance (from full to partial resistance: scores 1

to 3) or disease symptoms (from weak to full susceptibility: scores 4 to 6) and analysed by fitting a

proportional-odds model and performing an analysis of variance (a full statistical treatment of the

data is presented in Supplementary file 10). This analysis revealed significant differences between

groups of isolates ($\chi^2(6)=100$, p<0.001), and between rice genetic groups ($\chi^2(4)=161$, p<0.001), and a

significant interaction between these two variables ($\chi^2(24)=97$, p<0.001). The finding of a significant

interaction between groups of isolates (lineages or clusters) and rice types indicates that the effect of

the group of isolates on the proportion of compatible interactions differed between rice types,

suggesting adaptation to the host. This is also supported by the specific behaviour of certain lineages.

Isolates from lineage 2 had much lower symptom scores than the other lineages on all rice types

except temperate japonica, and the isolates of the *Yule* cluster were particularly virulent on indica

varieties (Figure 4A; Supplementary file 9). In comparisons of rice genetic groups, significantly higher

symptom scores were observed on temperate japonica rice than on the other types of rice, whereas

the varieties of the aromatic genetic group were significantly more resistant to rice blast (Figure 4B).

Together, these experiments therefore revealed significant differences in host range between

lineages, but this specialization to the host was not strict, because there was an overlap between

host ranges (Figure 4C-G).

Figure 4: Compatibility between 70 *P. oryzae* isolates and 44 rice varieties, representing five types of rice. Compatibility was

measured as symptom scores estimated using pathogenicity tests in controlled conditions. A: Symptom scores as a function

of the lineage of origin of isolates, or cluster of origin for isolates from lineage 1; B: Symptom scores as a function of the

type of rice; C-G: Symptom scores as a function of the lineage of origin of isolates, for each type of rice. Abbreviations: Y,

Yule; I, *International*; B, *Baoshan*; L, *Laos*; 2, lineage 2; 3, lineage 3; 4, lineage 4; Ind, indica; TeJ, temperate japonica; TrJ,

tropical japonica; Aro, aromatic. Small capitals indicate significant differences. All interactions between rice varieties and *P.*

oryzae isolates were assessed in three independent experiments, and the highest of the three symptom scores was used in

calculations. Boxen plots were drawn using function boxenplot() with PYTHON package SEABORN 0.11.1. Number of boxes was

controlled using option k_depth='trustworthy'. Grey horizontal lines represent the median.

Figure 4 – source data 1

Symptom scores for *P. oryzae* isolates inoculated onto rice varieties

Differences in the number and genetic variability of putative effector genes between *P. oryzae* lineages

Effectors are the most prominent class of fungal virulence factors. They act by manipulating host cellular pathways and are key determinants of the host range and fitness on compatible hosts (Liao *et al.* 2016; Yoshida *et al.* 2016). We therefore determined the differences in effector repertoires between *P. oryzae* lineages in terms of presence/absence and nucleotide polymorphisms, and analyzed whether this was a particularly dynamic fraction of the gene space. To this end, we used whole-genome sequencing data for 123 isolates, including 29 isolates sequenced in this study and 94 publicly available genomes (Gladieux *et al.* 2018a; Pordel *et al.* 2020; Zhong *et al.* 2018), 33 of which were also genotyped with our Infinium genotyping beadchip (Supplementary file 11; Appendix 2). Clustering analysis indicated that this dataset covered the four lineages of *P. oryzae*, and three of the clusters of lineage 1 (Laos, International and Baoshan; Appendix 2; Supplementary file 11). Assembly lengths for the 123 sequenced isolates ranged from 36.7 to 39.6 Mb, with a mean value of 38.1 Mb (Supplementary file 11). The number of assembled contigs longer than 500 bp ranged from 1010 to 2438, and the longest contig was 1.1 Mb long (Supplementary file 11).

Gene prediction identified 11,684 to 12,745 genes per genome, including 1,813 to 2,070 genes predicted to encode effector proteins. We found a significant effect of the lineage of origin on both the mean number of putative effector genes (ANOVA, F=5.54, p=0.0014), and the mean number of non-effector genes (ANOVA, F=7.95, p<0.001). Multiple comparisons revealed that mean numbers of putative effectors and non-effector genes were only significantly different between the genomes of lineage 2 and lineages 3 (Tukey's HSD test; p<0.001; Figure 5A; Figure 5B). An orthology analysis identified 14,573 groups of orthologous sequences (i.e. orthogroups), and we applied a rarefaction approach to the table of orthology relationships to estimate the size of accessory and core genomes in lineages while accounting for differences in sample size. For both putative effectors and the remaining of the gene space, we found that the number of core and accessory genes did not reach a plateau, instead displaying an almost linear relationship with the number of genomes resampled,

indicating that the number of core genes was probably much smaller, and the number of accessory

genes substantially higher than estimated from 123 genomes (Supplementary file 12). With

pseudosamples of size n=30 per lineage (i.e. excluding lineage 4, due to its small sample size), the

gene content was highly variable within lineages with only 61-71% of all predicted effectors, and 68-

73% of the remaining of the gene space being conserved in lineages 1-3 (Supplementary file 13).

Despite extensive variation of the gene content within lineages, the clustering of isolates based on

presence/absence variation for both putative effectors and the remaining of the gene space was

highly similar to that based on 3,868 SNPs (Figure 5D; Figure 5E), indicating that variation in gene

content and SNP allelic variation reflected similar genealogical processes.

For the identification of putative effectors potentially involved in the host specialization of

rice-infecting lineages of *P. oryzae*, we first identified orthogroups with different patterns of

presence/absence across lineages. Principal component analysis on presence/absence data identified

72 orthogroups of putative effectors accounting for 95% of the variance of the three principal

components differentiating the four lineages (Figure 5C; Figure 5D; Figure 5E), and potentially

contributing to the differences in host range between lineages.

Host specialization can also involve sequence divergence for effector proteins. We therefore

also scanned the corresponding genes for signatures of diversifying selection, i.e. an excess of non-

synonymous nucleotide diversity (ratio of non-synonymous to synonymous nucleotide diversity π_N/π_S

>1). We identified 185 orthogroups with $\pi_N/\pi_S > 1$ in at least one lineage, including 164 orthogroups

with $\pi_N/\pi_S > 1$ in only one lineage (lineage 1: 131 orthogroups; lineage 2: 18; lineage 3: 10; lineage 4:

5), and twelve, seven and two orthogroups with $\pi_N/\pi_S > 1$ in two, three or four lineages, respectively

(Supplementary file 14). None of these orthogroups corresponded to effectors previously

characterized as being involved in *P. oryzae* virulence on rice or other *Poaceae* hosts.

Figure 5. Pangenome analyses of 123 genomes of *P. oryzae*, representing four rice-infecting lineages. (A) and (B) Violin plots showing the number of putative effectors and non-effector proteins detected in each lineage, respectively; asterisks indicate significant differences in gene content (Tukey's HSD test; p<0.001). (C) and (D) principal components PC1 against PC2, and PC1 against PC6, respectively, in a principal component analysis of presence/absence of accessory putative effectors and non-effector proteins; between parentheses is the proportion of variance represented by each principal

component. (E) Presence and absence in lineage 1-4 of the 72 putative effectors with the highest contribution to principal
components 1, 2 and 6, in a principal component analysis of presence/absence data.

Figure 5 – source data 1

Size of core and accessory genomes for putative effectors and non-effector proteins

Figure 5 – source data 2

Presence and absence of putative effectors in *P. oryzae* genomes

Figure 5 – source data 3

Presence and absence of non-effector genes in *P. oryzae* genomes

Figure 5 – source data 4

Orthology table, with one orthogroup per line and one isolate per column

Geographic and climatic differentiation in the distribution of *P. oryzae* lineages

Our tests of adaptation to host and temperature under controlled conditions showed that

specialization was not strict, but it remained possible that fitness differences between lineages would

be sufficient under natural conditions to induce separation in different ecological niches, and/or that

our experimental conditions did not capture the full extent of the phenotypic differences. Here, we

tested the hypothesis that lineages thrive in geographically and/or environmentally different

conditions, which would provide indirect evidence for specialization to different habitats. We tested

this hypothesis by collecting geographic and climatic data for all isolates, using the GPS position of

the nearest city or the geographic center of their region of origin when the exact GPS position of

isolates was not available (Supplementary file 1). At a larger scale, clonal lineages 2 and 3 were both

widespread, with lineage 2 found on all continents, and lineage 3 on all continents except Europe.

Lineage 2 was the only lineage sampled in Europe (with the exception of a single isolate from lineage

1), whereas lineage 3 was more widespread in intertropical regions. Lineage 4 was mostly found in

South Asia (India, Bangladesh, Nepal), and in the USA and Africa (Benin, Tanzania). The recombining

lineage, lineage 1, was found on all continents, but mostly in Asia (79% of all isolates and 94% of all

genotypes). At a smaller scale, two, or even three different lineages were sampled in the same year in

the same city in 11 different countries from all continents. However, in only one instance did two

isolates from different lineages have identical GPS positions (isolates US0106 and US0107 sampled

from the same field). Thus, lineages had different, but partially overlapping distributions, including

some overlap with the distribution of the sexually reproducing lineage (lineage 1).

Figure 6. Geographic distribution of four lineages of *P. oryzae* and corresponding climatic data. (A, B) Pie charts

representing the distribution of the four lineages in the world (A) and in South, East and Southeast Asia (B), keeping only

isolates for which the sampling position was precisely known (i.e., for which the region, city or GPS position was

documented). Background map is from OpenStreetMap (CC-BY-SA) and represents the updated Köppen-Geiger climate

classification of main climates at world scale as described in (Kottek *et al.* 2006). (C, D) Outlying Mean Index analysis, testing

the separation of ecological niches among lineages, with (C) canonical weights of the 19 environmental variables included in

the analysis, and (D) site coordinates (dots) and realized niches of lineages represented as ellipses. Variable bio13 co-

localizes with bio16, and variables bio1 and bio3 co-localize with bio11. The 11 temperate variables included in the analysis

are listed in the Methods section.

Figure 6 – source data 1

GPS position of isolates

We then investigated whether differences in the geographic range of lineages were

associated with differences in climatic variables. By plotting sampling locations onto a map of the

major climate regions (Kottek *et al.* 2006) we were able to identify clear differences in the climatic

distributions of lineages 2 and 3, with lineage 2 mostly found in warm temperate climates and lineage

3 found in equatorial climates (Figure 6; Source code 1). We used the outlying mean index (OMI),

which measures the distance between the mean habitat conditions used by a lineage and the mean

habitat conditions used by the entire species, to test the hypothesis that different lineages are

distributed in regions with different climates. Using information for 19 climatic variables (biomes)

from the WorldClim bioclimatic variables database (Fick & Hijmans 2017) for all sampling locations,

we obtained statistically significant results, in a permutation test on OMI values, for lineages 2, 3 and

4 (Source code 2; permutation test: lineage 1: OMI=2.14, *p*=0.620; lineage 2: OMI=6.54, *p*<0.001;

lineage 3: OMI=2.08, *p*<0.001; lineage 4: OMI=13.73, *p*=0.017). The OMI analysis, in which the first

two axes accounted for 69% and 25% of the variability, respectively, revealed that lineage 2 was more

frequent in regions with a high annual temperature range (biome 7) or a high degree of seasonality

(biome 4); lineage 4 was associated with regions with high levels of seasonal precipitation (biomes

13, 16 and 18), and lineage 3 was more frequent in regions with high temperatures (biomes 1, 6, 10

and 11) and high levels of isothermality (biome 3), characteristic of tropical climes (Source code 2;

Figure 6; Supplementary file 15).

DISCUSSION

This study provides a detailed overview of the worldwide distribution of genetic and phenotypic

diversity in *P. oryzae*, and of the eco-evolutionary factors underlying the emergence and maintenance

of multiple divergent lineages in this widespread pathogen. We detected three clonal lineages and

one recombining lineage with broad and largely overlapping distributions, corresponding to the four

groups previously identified on the basis of fewer markers or fewer isolates (Gladieux *et al.* 2018c;

Latorre *et al.* 2020; Saleh *et al.* 2014; Tharreau *et al.* 2009; Zhong *et al.* 2018). Our findings provide

strong support for previous findings, and extend these findings by revealing that the lack of

recombination and recent admixture in widespread lineages persists with deeper sampling, and by

providing evidence for the subdivision of the recombining lineage into three Asian clusters and one

international cluster. This deep sampling of the natural genetic variationalso reveals that, despite

distributions displaying a broad overlap over large scales, rice blast lineages thrive in areas differing in

terms of the prevailing environmental conditions and types of rice grown. We also provide

experimental evidence that host-specialization, female sterility and early post-mating genetic

incompatibilities act as strong barriers to gene flow between lineages. These results highlight the

ability of pathogen spread over heterogeneous habitats and of divergent populations of a crop

species to create niche separation and barriers to gene flow between different invasive lineages.

Niche separation and barriers to gene flow between clonal lineages

Many invasive plant pathogens consist of multiple divergent lineages coexisting on the same host,

over large areas, raising questions about the maintenance of different lineages in the face of potential

gene flow and competitive exclusion. We show that female sterility and intrinsic genetic

incompatibilities represented strong barriers to gene flow between the clonal lineages, potentially

accounting for their maintenance over time, without fusion, despite the compatibility of mating types

(at least between lineages 2 and 4 [*Mat1.1*] on the one hand, and lineage 3 [*Mat1.2*] on the other).

These barriers to gene flow may have contributed to the establishment of lineage specialization

(because reproductive isolation facilitates adaptation by limiting the flow of detrimental ancestral

alleles in populations adapting to a new habitat), accounting for the maintenance of the clonal

lineages on the same host species, when faced with competitive exclusion. Our analyses also show

that, despite widely overlapping large-scale distributions, the different lineages are essentially found

in different regions, with different climatic characteristics, when their distributions are observed at a

finer scale. Lineage 1 was mostly sampled in Southeast Asia, lineage 4 predominated in India, and

lineages 2 and 3 had global distributions, but the geographic ranges of all these lineages overlapped.

However, an analysis of climatic data indicated that lineage 2 predominated in temperate climates,

where temperate japonica rice is grown, lineage 3 predominated in tropical climates, in which indica

rice is grown, and lineage 4 predominated in regions with high levels of seasonal precipitation in

which the indica and aromatic rice types are the principal rice types grown. Despite the finding of

separation in different climatic regions, our experiments revealed no strong differences between

lineages in terms of sporulation and mycelial growth on synthetic media at different temperatures.

This suggests that, if adaptation to temperature occurs in this pathogen, it was not measured by our

experiments, either because it involves traits other than sporulation and hyphal growth, or because

the *in vitro* conditions were not suitable to demonstrate differences. The host range varied across

lineages, but all host ranges overlapped, indicating that host specialization was not strict, or was not

fully captured by our experiments. Nevertheless, adaptation to the host or small differences in

temperature optima may, nonetheless, further increase niche separation and reduce gene flow

between lineages, via pre-mating barriers, such as immigrant non-viability (i.e., lower rates of contact

between potential mates due to the mortality of immigrants; (Gladieux *et al.* 2011a, b; Nosil *et al.*

2005)), and post-mating barriers, such as ecological hybrid non-viability (i.e., lower survival of ill-

adapted hybrid offspring).

Pestification can foster lineage divergence

There is a direct connection between disease emergence and speciation, because the emergence of

new pathogens can be facilitated, caused by or associated with strong divergent selection in

organisms mating within or on their hosts, and by major changes in demographics and geographic

isolation (Giraud *et al.* 2006; Hansen 1987; Restrepo *et al.* 2014; Stukenbrock 2013). There is also a

connection between disease emergence and mode of reproduction, because disease emergence

should be facilitated by multiple asexual cycles, corresponding to multiple cycles of selection for local

adaptation without migration introducing locally deleterious immigrant alleles and recombination

breaking down locally advantageous allelic combinations (Giraud *et al.* 2006). However, theoretically,

in the long term, asexual organisms should accumulate a greater load of deleterious mutations and

be less able to fix advantageous mutations, resulting in higher rates of extinction and lower rates of

speciation (Orr 2000; Otto & Lenormand 2002). The pattern of population structure revealed in *P.*

oryzae is seems to be intermediate between observations for invasive fungi with fully clonal

population structures — such as the pathogens causing wheat brown rust (Fellers *et al.* 2020),

boxwood blight (LeBlanc *et al.* 2020) or verticillium wilt (Short *et al.* 2015; Short *et al.* 2014) — and

invasive fungi with barriers maintaining invasive lineages that are permeable and allow hybridization

and introgression to occur, such as the pathogens causing *Dothistroma* needle blight (Capron *et al.*

2021), Dutch elm disease (Hessenauer *et al.* 2020), soybean anthracnose *Colletotrichum truncatum*

(Rogério *et al.* 2021), and coffee rust (Ramírez-Camejo *et al.* 2021; Silva *et al.* 2018). The global

structure of *P. oryzae* is more reminiscent of that of the pathogens causing chestnut blight (Dutech *et*

al. 2012; Stauber *et al.* 2021), wheat yellow rust (Ali *et al.* 2014; Bueno-Sancho *et al.* 2017), and hop

powdery mildew (Weldon *et al.* 2021), for which clonal invasive lineages coexist with recombining

lineages, although the proximal causes for the loss of sexual reproduction may differ between these

models and rice blast. Another marked difference from most of the abovementioned pathogens is

that the clonal lineages of rice blast diverged about a thousand years ago (Gladieux *et al.* 2018c;

Latorre *et al.* 2020; Zhong *et al.* 2018), corresponding to several thousand asexual generations.

Recombining lineage 1 may, at some point, become a source of adaptive introgression, reinvigorating

the genomes of clonal lineages to reduce the mutational load, introducing beneficial variation, and

leading to marked changes in population structure. Thus, although our work shows that pestification

can drive the emergence of reproductive barriers, we cannot predict whether these lineages will

continue to exist in their current form for the foreseeable future.

Loss of female sterility as the proximal cause of shifts to clonality.

The loss of sexual reproduction, caused by demographic events or selection against sex, is a frequent

change in the life history traits of invasive fungal pathogens (Gladieux *et al.* 2015; Hessenauer *et al.*

2021). However, the underlying changes and driving factors remain largely unexplored. In *P. oryzae*,

the release of constraints maintaining sex in the short term, such as the need for sexual structures

for winter survival or long-distance dispersal, is probably not responsible for the loss of sex, because

the likelihood of overwintering on seeds and the rate of human-mediated transportation would not

be expected to differ between clonal lineages 2-4 and recombining lineage 1. The protection of

locally advantageous allelic combinations is a more plausible explanation for the loss of sex in

lineages 2-4, because, as shown here, these lineages are found in, and adapted to, different hosts

and environmental conditions. This leaves open the question of the maintenance of sex in lineage 1

open, but it is possible the diversity of cultivated and wild rice diversity of rice is higher in the

geographical range of this lineage (Saleh *et al.* 2012b). It is difficult to draw firm conclusions about

the evolutionary causes for selection against sexual reproduction, but the finding of an

internationally distributed cluster in lineage 1 sheds light on the possible sequence of proximal

events underlying the emergence of clonal lineages in this pathogen. The *International* cluster of

lineage 1 displayed genetic footprints of recombination and sexual reproduction (in the form of an

excess of homoplastic variants or relatively balanced mating types), but the signal of recombination

detected here may be purely historical, because we also found that the clonal fraction and the

frequency of sterile females were very high in this cluster. The loss of female fertility may, therefore,

be an early and major cause of the shift towards asexual reproduction. Some genotypes assigned to

the international cluster of lineage 1 may, in the future, come to resemble other clonal lineages in

analyses of genealogical relationships, and form a long branch stemming from the central lineage 1,

particularly for genotypes already displaying clonal propagation, such as genotypes 2, 18, 58, 98 and

254. However, according to the principle of competitive exclusion, competition between clonal

lineages should only allow hegemonic expansion of the most competitive clonal groups, unless they

separate into different ecological niches. The asexual spread of the International cluster in a separate

ecological niche may ultimately lead to the fixation of a single mating type, as in clonal lineages 2-4.

This model could also apply to other fungal pathogens with similar life history strategies.

Differences in effector repertoires between lineages

Effector repertoires play a key role in pathogen specialization, in fungal pathogens in general

(Schulze-Lefert & Panstruga 2011b), and *P. oryzae* in particular (Asuke *et al.* 2020; Couch *et al.* 2005;

Liao *et al.* 2016; Sweigard *et al.* 1995; Takabayashi *et al.* 2002). Previous work suggested that clonal

lineages of *P. oryzae* possess smaller effector repertoires (Latorre *et al.* 2020) and that the clonal

lineage associated with japonica rice (lineage 2) possesses more effectors, and, in particular, Avr-

effectors, than other clonal lineages (Latorre *et al.* 2020; Liao *et al.* 2016). Our analyzes on a larger

set of putative effectors (ca. 2000 in our study, vs. 13 Avr-effectors in (Liao *et al.* 2016) and 178

known and candidate effectors in (Latorre *et al.* 2020)) confirm this trend, because lineage 2

associated with temperate japonica has significantly more putative effectors than lineage 3

associated with indica. However, the number of non-effector genes is also greater in lineage 2 than in

lineage 3, and thus it remains possible that the larger effector complement of lineage 2 is a mere

consequence of a larger number of genes. We also found that patterns of presence / absence

variation mirrored patterns of population subdivision based on SNP allelic variation, and, thus, that

the differential sorting of both nucleotide polymorphism and gene content across lineages reflects

similar genealogical processes. Our multivariate analyses identified 72 effectors making the greatest

contribution to the differentiation of the four lineages in terms of presence / absence. It is possible

that the frequency differences for these 72 effectors were due to chance events during bottlenecks

at the onset of lineage formation, or, alternatively, that their differential loss was important for the

initial adaptation of lineages to new rice populations or subspecies. Interestingly, two of these 72

effectors, AvrRmg8 (=OG0011611) and PWL3 (=OG0011928), are host range determinants in wheat-

infecting *P. oryzae* because they trigger immunity in wheat varieties carrying the resistance genes

Rmg8 (Anh *et al.* 2018) or *PWT3* (Inoue *et al.* 2017; Kang *et al.* 1995). Nucleotide diversity at

synonymous and non-synonymous sites also identified 185 effectors with signatures of diversifying

selection in one or more lineages, possibly mediated by coevolutionary interactions with host

molecules. In the future, it will be interesting to determine the molecular targets of these effectors

and to decipher the relationship between their polymorphism and their mode of action.

Lineage 1 as a threat to global rice production

Our results also indicate that lineage 1 may pose a major threat to rice production. Most of the

effectors with signatures of diversifying selection were identified in lineage 1, highlighting the greater

ability of this lineage to fix advantageous mutations rapidly. Eleven of the 12 most multivirulent

isolates (lesion type > 2 on more than 40 varieties tested) belonged to lineage 1, and the *Yule* cluster,

in particular, was highly pathogenic on indica varieties. The propagation of such highly pathogenic

genotypes, belonging to a lineage that has both mating types and fertile females, should be

monitored closely. The monitoring of lineage 1 is all the more critical because this lineage has an

intermediate geoclimatic distribution that overlaps largely with the distributions of the other three

lineages, and some of the attempted crosses between clonal lineage 2-4 isolates and recombining

lineage 1 isolates produced viable progeny. This finding confirms the possibility of gene flow into this

lineage, as previously demonstrated on the basis of admixture mapping (Gladieux *et al.* 2018b;

Latorre *et al.* 2020), but it also raises the question of the threat posed by gene flow from lineage 1,

which is highly pathogenic, to the other lineages.

Concluding remarks

Our study of genetic and phenotypic variation within and between clonal lineages, and within the

recombinant lineage of *P. oryzae* suggests a scenario for the emergence of widespread clonal

lineages. The loss of female fertility may be a potent driver of the emergence of asexually

reproducing groups of clones. The reproductive isolation generated by the loss of sex and the

accumulation of mutations due to the absence of sexual purging would facilitate the specialization of

some of these clonal groups, leading to the competitive exclusion of the least efficient clonal groups,

and, finally, to the propagation of clonal lineages fixed for a single mating type. Our results, thus,

demonstrate that the spread of a pathogen across heterogeneous habitats and divergent populations

of a crop species can lead to niche separation and reproductive isolation between different invasive

lineages in the pathogen.

MATERIALS AND METHODS

Biological material

We chose 886 *P. oryzae* isolates collected on Asian rice between 1954 and 2014 as representative of

the global genetic diversity of the fungus. Isolates were selected on the basis of microsatellite data, to

maximize the number of multilocus genotypes represented ((Saleh *et al.* 2014) and unpublished

data), or based on geographic origin in the absence of genotypic data, to maximize the number of

countries represented in the dataset.

Sixty-eight isolates were selected for experimental measurements of reproductive success,

adaptation to host, and growth and sporulation experiments at different temperatures

(Supplementary file 1). This subset of isolates included 10 isolates from each of the three clonal

lineages (lineages 2-4), 27 isolates from the various clusters within lineage 1 [*Baoshan* (9 isolates),

International (10), *Laos* (8), and *Yule* (11)].

Forty-four varieties were chosen as representative of the five main genetic subgroups of

Asian rice (Garris *et al.* 2005): indica (Chau, Chiem chanh, DA11, De abril, IR8, JC120, Pappaku), aus

(Arc 10177, Baran boro, Black gora, DA8, Dholi boro, Dular, FR13 A, JC148, Jhona 26, Jhona 149,

Kalamkati, T1, Tchampa, Tepi boro), temperate japonica (Aichi asahi, Kaw luyoeng, Leung pratew,

Maratelli, Nep hoa vang, Nipponbare, Sariceltik, Som Cau 70A), tropical japonica (Azucena,

Binulawan, Canella de ferro, Dholi boro, Gogo lempuk, Gotak gatik, Trembese) and aromatic (Arc

10497, Basmati lamo, Dom zard, Firooz, JC1, Kaukkyisaw, N12). The Maratelli (temperate japonica)

and CO39 (indica) varieties were used as susceptible controls.

Genotyping

Pyricularia oryzae isolates were genotyped at 5,657 genomic positions with an Illumina Infinium

beadchip microarray carrying single-nucleotide polymorphisms identified in 49 previously

characterized genomes of rice- and barley-infecting *P. oryzae* isolates (Gladieux et al., 2018b). The

final dataset included 3,686 biallelic SNPs without missing data.

Whole genome sequencing

We used a set of 123 whole-genome sequences to investigate differences in effector content

between clusters and lineages, including 94 publicly available genomes and 29 new genomes that we

selected to increase sample size for the International and Laos clusters of lineage 1, and for lineage 4

(Supplementary file 11). The 29 isolates were sequenced using Illumina HiSeq 3000. Isolates were

grown on rice flour-agar medium for mycelium regeneration, then in liquid rice flour medium (Adreit

et al. 2007). Genomic DNA extraction was carried out with >100 mg of fresh mycelium from liquid

culture. Fresh mycelium dried on Miracloth paper was crushed in liquid nitrogen. Nucleic acids were

subsequently extracted using an extraction buffer (2 % CTAB – 1.4 M NaCl – 0.1 M Tris-HCl pH 8 – 20

mM EDTA pH 8 added before use with a final concentration of 1 % Na_2SO_3), then purified with a

chloroform:isoamyl alcohol (24:1) treatment, precipitated overnight in isopropanol, and rinsed with

70% ethanol. The extracted nucleic acids were further treated with RNase A (0.2 mg/mL final

concentration) and purified with another chloroform:isoamyl alcohol (24:1) treatment followed by

overnight precipitation in ethanol. The concentration of extracted genomic DNA was assessed on

Qubit® using the dsDNA HS Assay Kit. The purity of the extracted DNA was checked by verifying that

the 260/280 and 260/230 absorbance ratios measured with a NanoDrop spectrophotometer were

between 1.8 and 2.0. Preparation of TruSeq nano library preparation and HiSeq3000 sequencing (150

nucleotide reads, and 500 bp insert size) were performed at GeT-PlaGe (INRAE, Toulouse, France).

Genome assembly, gene prediction, orthology analysis, effector prediction, and summary

statistics of nucleotide variation

For the 123 sequenced isolates included in the dataset, low-quality reads were removed with

CUTADAPT software (Martin 2011). Reads were assembled with ABYSS 2.2.3 (Jackman *et al.* 2017;

Simpson *et al.* 2009), using different K-mer sizes. For each isolate, we chose the assembled sequence

with the highest N50 for further analyses. Genes were predicted with BRAKER 2.1.5 (Hoff *et al.* 2015;

Hoff *et al.* 2019) using RNAseq data (Pordel *et al.* 2020) as extrinsic evidence for model refinement.

Genes were also predicted with AUGUSTUS 3.4.0 (Stanke & Morgenstern 2005)(training

set=*Magnaporthe grisea*) and gene models that did not overlap with the gene models identified with

BRAKER were added to the GFF file generated with the latter. Repeated regions were masked with

REPEATMASKER 4.1.0 (http://www.repeatmasker.org/). The completeness of genome assemblies and

gene predictions was assessed with BUSCO (Simão *et al.* 2015), and completeness in BUSCO genes

was greater than 93.7% (Supplementary file 11). Putative effector genes were identified as genes

encoding proteins predicted to be secreted by at least two of three methods [SIGNALP 4.1 (Nielsen

2017), TARGETP (Emanuelsson *et al.* 2000) and PHOBIUS (Käll *et al.* 2004)], with no predicted

transmembrane domain based on TMHMM analysis (Krogh *et al.* 2001), no predicted motif of

retention in the endoplasmic reticulum based on PS-SCAN (Bhagwat & Aravind 2007), and no CAZy

annotation based on DBSCAN v7 (Yin *et al.* 2012). Differences in the numbers of putative effectors and

non-effectors between lineages (ANOVA and Tukey's HSD test) were assessed with the SCIPY 1.6.0

package in PYTHON. Homology relationships betweenpredicted genes were established with

ORTHOFINDER v2.4.0 (Emms & Kelly 2015). Principal component analysis of presence/absence

variation for putative effectors and non-effector proteins was performed with the R package PRCOMP.

We used the get_pca_var() in PRCOMP to extract the results for variables (i.e. effectors) and to identify

the effectors making the largest contributions to the principal components, defined as effectors with

a contribution to loadings of PC1, PC2 and PC6 greater than 1%. For estimates of the numbers of core

and accessory genes, we used a rarefaction approach to account for differences in sample size across

lineages. For each pseudo-sample size, we estimated the size of the core and accessory genome in

each lineage using a maximum of 2000 pseudosamples. Sequences for each orthogroup were aligned

and cleaned with TRANSLATORX (Abascal *et al.* 2010), using default parameters. The ratio of non-

synonymous to synonymous diversity π_N/π_S was calculated for each orthogroup, for lineages with a

sample size of at least four, with EGGLIB 3 (https://www.egglib.org).

Population subdivision and recombination

The dataset used for population genetic analysis had no missing data. Clones were therefore

identified as isolates with identical genotypes at 3,686 sites. Among the 886 *P. oryzae* isolates

genotyped, we identified 264 different multilocus genotypes (i.e., "clones"), which were used for the

analysis of population subdivision. We used the sNMF program to infer individual ancestry

coefficients in *K* ancestral populations. This program is optimized for the analysis of large datasets

and does not assume Hardy-Weinberg equilibrium. It is, therefore, more appropriate to deal with

inbred or clonal lineages (Frichot *et al.* 2014) . We used SPLITSTREE version 4 (Huson & Bryant 2006) to

visualize relationships between genotypes in a phylogenetic network, with reticulations to represent

the conflicting phylogenetic signals caused by homoplasy. PCA was performed as implemented in the

PYTHON package SCIKIT-ALLEL v. 1.3.3. (https://zenodo.org/record/4759368#.YL4R5DaiHwQ). Weir and

Cockerham's F_{ST} was calculated with the R package HIERFSTAT, by the WC84 method (Goudet 2005).

We tested the null hypothesis of clonality using the pairwise homoplasy index (PHI; (Bruen *et*

al. 2006)), maximum χ^2 (Max χ^2, (Maynard Smith 1992)) and neighbour similarity score (NSS;

(Jakobsen & Easteal 1996)), as implemented in the PHIPACK program

(https://www.maths.otago.ac.nz/~dbryant/software.html). Homoplastic sites have sequence

identities that are not inherited from a common ancestor, but instead derived from independent

events in different branches, such as recurrent mutations, sequencing errors or recombination.

Homoplastic sites were identified by concatenating genotypes at all sites, inferring a genealogy with

the GTR+GAMMA model in RAxML v.8.2.12 (Stamatakis 2014), and mapping mutations onto the

genealogy with the 'Trace All Characters' function of MESQUITE under maximum parsimony (Maddison

& Maddison 2019). In the resulting matrix of ancestral states for all nodes, the number of mutations

occurring at each site was counted and sites displaying multiple mutations across the genealogy were

considered to be homoplastic. We used POPLDDECAY (Zhang *et al.* 2019) to measure linkage

disequilibrium (r^2) as a function of distance between pairs of SNPs. POPLDDECAY was configured with a

maximum distance between SNPs of 400 kbp, and a minor allele frequency of 0.005.

Experimental measurement of reproductive success and female fertility

Pyricularia oryzae is a heterothallic fungus with two mating types (*Mat1.1* and *Mat1.2*). Sexual

reproduction between strains of opposite mating type can be observed in laboratory conditions, and

results in the production of ascospores within female sexual structures called perithecia (Saleh *et al.*

2012a). Crosses were carried out on a rice flour agar medium (20 g rice flour, 2.5 g yeast extract, 15 g

agar and 1 L water, supplemented with 500 000 IU penicillin G after autoclaving for 20 min at 120°C),

as described by (Saleh *et al.* 2012a). We assessed reproductive success by determining the

production of perithecia produced after three weeks of culture at 20°C under continuous light. Each

cross was performed twice and we determined the mean number of perithecia across repeats. We

further assessed the presence of asci and germinating ascospores in perithecia for a subset of

crosses, by excising perithecia with a scalpel and counting the number of germinated filaments for

each individual ascus after overnight incubation on water agar. The subset of crosses included 10

Mat1.1 isolates (lineage 1: CH0999, CH1065, CH1076; lineage 2: CH0092, MC0016, SP0006; lineage 4:

IN0017, IN0092, NP0070, CH0718) and 6 Mat1.2 isolates (lineage 1: CH0997; CH1083, CH1120;

lineage 3: BR0019, CH0549, MD0929).

We measured female fertility for 210 isolates (listed in **Supplementary file 1**), by monitoring

perithecium production in crosses involving the tester strains CH0997 (Mat1.2) and CH0999 (Mat1.1).

On synthetic media, perithecia are formed at the zone of contact between parental mycelia. Isolates

forming perithecia are described as female-fertile. Perithecia can be formed by the two interacting

partners or by one partner only. If the two partners are female-fertile, two parallel lines of perithecia

are observed at the contact zone between mycelia. If only one partner is female-fertile, only one line

of perithecia is observed, at the edge of the mycelium of the fertile isolate.

Pathogenicity tests

Compatibility between *P. oryzae* isolates and rice plants from the five main genetic subgroups of rice

(indica, temperate japonica, tropical japonica, aus and aromatic) was measured in controlled

conditions. We inoculated on 46 varieties (see section Biological Material), with 70 isolates.

Inoculations were performed as described by (Gallet *et al.* 2016). Conidial suspensions (25 000

conidia.mL^{-1}) in 0.5% gelatin were sprayed onto three-week-old rice seedlings (> 6 plants/variety).

The inoculated plants were incubated for 16 hours in the dark at 27°C and 100% humidity and then

for seven days with a day/night alternation (13 hours at 27°C/11 hours at 21°C), before the scoring of

symptoms. Lesion type was rated from 1 to 6 (Gallet *et al.* 2016) and the symptom type was assessed

visually on leaves. Each interaction was assessed in three independent experiments. Symptoms

scores are ordinal-scale variables (i.e. variables that can be ranked but are not evenly spaced). We

therefore analyzed the data with a proportional-odds model with the clm() function implemented in

the ORDINAL version 2018.8-25 in R, including rice type and pathogen lineage as main effects, and the

rice type – pathogen lineage interaction. A comprehensive mixed model including all experimental

replicates, with isolate as a random effect, was also tested, but did not display convergence. We

therefore used the maximum of the three scores in calculations, as this is what most accurately

represented the ability of an isolate to exploit a given host genotype. Analyses were also conducted

with median values, and gave similar results (not shown). We used an analysis of deviance to

evaluate whether effect and interaction terms were statistically significant. Pairwise comparisons of

significant factors were performed after computing least-squares means with LSMEANS version 2.30-0

in R, with Tukey adjustment. The R commands and their outputs are provided in Supplementary file

9.

Mycelial growth and sporulation rate at different temperatures

Mycelial growth and sporulation rate were measured for 41 isolates at five different temperatures (10°C, 15°C, 20°C, 25°C and 30°C). For each isolate, Petri dishes containing PDA medium were inoculated with mycelial plugs placed at the center, and incubated in a growth chamber at a fixed temperature. Mycelium diameter was estimated as the mean of two measurements made along two perpendicular axes at different time points. At the end of the experiment, conidia were collected by adding 5 mL of water supplemented with 0.01% Tween 20 to the Petri dish and rubbing the surface of the mycelium. Conidia were counted with a hemocytometer. Three or four independent experiments were performed for each isolate, at each temperature. We had initially planned to carry out only three independent experiments. However, a fourth experiment was eventually performed for some temperature conditions because some isolates did not grow in the first experiment, and because some cultures were invaded by mites and had to be discarded in the second and third experiments.

Mycelium growth was analyzed independently for each temperature with linear mixed-effects models. The variable used in statistical analysis was the square root of the mean mycelial diameter measured along two perpendicular axes on a Petri plate. Post-hoc pairwise comparisons were performed after computing least-squares means with LSMEANS version 2.30-0 in R, with Tukey adjustment. The R commands and their outputs are provided in Supplementary file 4.

Sporulation rates were analyzed with a Kruskal-Wallis test with RSTATIX version 0.7.0 in R. The variable used in statistical analysis was the median number of spores calculated across replicates. Post-hoc pairwise comparisons were performed with Dunn's non-parametric multiple comparison test. The R commands and their outputs are provided in Supplementary file 7.

Statistical analyses of climatic data

The outlying mean index (OMI), or marginality, is used to study niche separation and niche breadth. The OMI approach evaluates whether a particular lineage is mostly associated with some climatic conditions or can be sampled with the same probability throughout the range of the species. The

approach is based on a multivariate analysis (here a PCA) of climatic data associated with the

locations included in the dataset, regardless of the identity of the lineages sampled or the number of

isolates sampled. An index of marginality or specialization (the outlying mean index) is then

calculated for each lineage represented in the dataset. This index measures the distance between

the average habitat of a given lineage and the average habitat conditions of the area studied

(corresponding to the distribution of a hypothetical species uniformly distributed under all

conditions). Lineages are then placed in environmental conditions that maximize their OMIs, and

permutation tests are used to test whether the distribution of a lineage differs from random. OMI

gives the same weight to all sampling locations, whether rich or poor in individuals or lineages, and is

particularly suitable in cases in which sampling is not homogeneous. Environmental values, consisting

of 19 biome values (WorldClim bioclimatic variables (Fick & Hijmans 2017)), were retrieved for all

sampling locations. The 11 temperature variables included in the analysis were (bio1) annual mean

temperature, (bio2) mean diurnal range, (bio3) isothermality [100*(bio2/bio7)], (bio4) temperature

seasonality [standard deviation*100], (bio5) maximum temperature of the warmest month, (bio6)

minimum temperature of the coldest month, (bio7) annual temperature range [bio5-bio6], (bio8)

mean temperature of the wettest quarter, (bio9) mean temperature of the driest quarter, (bio10)

mean temperature of the warmest quarter, (bio11) mean temperature of the coldest quarter. The

eight precipitation variables included in the analysis were (bio12) annual precipitation, (bio13)

precipitation of the wettest month, (bio14) precipitation of the driest month, (bio15) precipitation

seasonality (coefficient of variation), (bio16) precipitation of the wettest quarter, (bio17)

precipitation of the driest quarter, (bio18) precipitation of the warmest quarter, (bio19) precipitation

of the coldest quarter. The 11 climatic variables were normalized before analysis. A contingency table

was generated to associate the number of isolates from each lineage with each sampling location.

Only isolates from our dataset with a precisely known geographic origin (region, city or GPS position)

were included in the analysis. A random permutation test with 10,000 permutations was used to

assess the statistical significance of marginality for each lineage.

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1878.

1123

1124 **DATA AVAILABILITY**

1125 All raw sequencing data are deposited under accession codes PRJEB42377 and PRJEB46618.

1126

The following data sets were generated:

1. European Nucleotide Archive

M Thierry, F Charriat, J Milazzo, H Adreit, S Ravel, S Cros-Arteil, S Borron, V Sella, T Kroj, R Ioos, E Fournier, D Tharreau, P

Gladieux

ID PRJEB42377 Whole genome sequencing of Pyricularia oryzae fungi isolated from rice.

2. European Nucleotide Archive

M Thierry, F Charriat, J Milazzo, H Adreit, S Ravel, S Cros-Arteil, S Borron, V Sella, T Kroj, R Ioos, E Fournier, D Tharreau, P

Gladieux

ID PRJEB46618 Whole genome sequencing of Pyricularia oryzae fungi from Benin and Burkina Faso

3. Zenodo

M Thierry, F Charriat, J Milazzo, H Adreit, S Ravel, S Cros-Arteil, S Borron, V Sella, T Kroj, R Ioos, E Fournier, D Tharreau, P

Gladieux

Single-nucleotide polymorphisms, genome assemblies, genome annotations, and gene predictions of Pyricularia oryzae

isolates from rice

doi 10.5281/zenodo.4561581

4. Zenodo

M Thierry, F Charriat, J Milazzo, H Adreit, S Ravel, S Cros-Arteil, S Borron, V Sella, T Kroj, R Ioos, E Fournier, D Tharreau, P

Gladieux

Single-nucleotide polymorphisms in isolates Pyricularia oryzae isolates from rice and other hosts

doi: 10.5281/zenodo.5500197

1147

The following previously published data sets were used:

1. European Nucleotide Archive

A Pordel, S Ravel, F Charriat, P Gladieux, S Cros-Arteil, J Milazzo, H Adreit, M Javan-Nikkhah, A Mirzadi-Gohari, A Moumeni,
D Tharreau
ID PRJEB41186 Origin and evolutionary history of Pyricularia oryzae fungal pathogens infecting maize and barnyard grass in
Iran
1154
1155 2. European Nucleotide Archive

1156 Z Zhong, M Chen, L Lin, Y Han, J Bao, W Tang, L Lin, Y Lin, R Somai, L Lu, W Zhang, J Chen, Y Hong, X Chen, B Wang, WC Shen,
1157 G Lu, J Norvienyeku, DJ Ebbole, Z Wang
1158 ID PRJNA354675 Population Genomic Analysis of the Rice Blast Fungus Reveals Specific Events Associated With Expansion
1159 of Three Main Clades
1160
1161 3. National Center for Biotechnology Information

1162 ID PRJNA320483 Genome-based Molecular Diagnostics for the Blast Fungus

1163
1164 4. National Center for Biotechnology Information

1165 ID PRJNA417903 Resequencing of a global collection of *Pyricularia oryzae* isolates
1166
1167

ADDITIONAL FILES

Source code 1. R code used to prepare maps in Figure 6.

Source code 2. R code used for analyses based on the Outlying Mean Index, including Principal Component
Analysis presented in Figure 6.

Supplementary file 1. Isolates of Pyricularia oryzae and analysis of genotyping technical replicates

Supplementary file 2. Assignment of genotypes to clusters identified in this study, in Gladieux et al. 2018b (A)
and in Saleh et al. 2014 (B and C).

Supplementary file 3. Linkage disequilibrium as a function of physical distance.

Supplementary file 4. F_{ST} estimates between clusters within lineage 1.

Supplementary file 5. Analysis of mycelial growth rates.

Supplementary file 6. Mycelial growth rates.

Supplementary file 7. Analysis of sporulation rates.

Supplementary file 8. Sporulation rates.

Supplementary file 9. Matrix of compatibility between four lineages of *P. oryzae* isolates and five subspecies of
rice plants, as determined by pathogenicity tests in controlled conditions. Numbers in the matrix represent the
maximum symptom score across three replicate experiments. Numbers at the right and bottom margins
represent the number of compatible interactions observed (symptom score >2) for varieties and isolates,
respectively.

Supplementary file 10. Analysis of pathogenicity data.

Supplementary file 11. Sequenced isolates.

Supplementary file 12. Estimating the size of the core and accessory genome using a rarefaction approach. For
each lineage, genomes were resampled in <=2000 combinations of N-1 genomes (N being the sample size). (A)
Non-effector genes. (B) Putative effectors.

Supplementary file 13. Size of core and accessory genomes for non-effector proteins (A) and putative effectors (B) estimated using a rarefaction approach with a pseudo-sample size of n=30 genomes. Lineage 4 was not included due to small sample size.

Supplementary file 14. Summary statistics of non-synonymous and synonymous polymorphism in four lineages of *P. oryzae.* For each lineage, only orthogroups with sample size >=4 were included in calculations. Lineage: lineage of origin of genomes used in calculations for each orthogroup. Orthogroup: orthogroup ID. N: sample size. numNS: number of non-synonymous sites. numS: number of synonymous sites. SNS: number of non-synonymous segregating sites. SS: number of synonymous segregating sites. PiNS: non-synonymous nucleotide diversity. PiS: synonymous nucleotide diversity.

Supplementary file 15. ecological niches of the four major lineages (referred to as L1 to L4) considering each of the 19 biomes individually (referred to as Bio1 to Bio19). The x-axis represents the outlying mean index (OMI), which measures the distance between the mean habitat conditions used by a lineage and the mean habitat conditions used by the entire species, to test the hypothesis that different lineages are distributed in regions with different climates.