# Evolution within a given virulence phenotype (pathotype) is driven by changes in aggressiveness: a case study of French wheat leaf rust populations

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**ABSTRACT**

Plant pathogens are constantly evolving and adapting to their environment, including their host. Virulence alleles emerge, and then increase, and sometimes decrease in frequency within pathogen populations in response to the fluctuating selection pressures imposed by the deployment of resistance genes. In some cases, these strong selection pressures cannot fully explain the evolution observed in pathogen populations. A previous study on the French population of *Puccinia triticina*, the causal agent of wheat leaf rust, showed that two major pathotypes — groups of isolates with the same combinations of virulences — predominated but then declined over the 2005-2016 period. The relative dynamics of these two pathotypes — 166 317 0 and 106 314 0 — relative to the others present in the population could not be explained solely by the frequency of *Lr* genes in the landscape. Within these two pathotypes, we identified two main genotypes that emerged in succession. We assessed three components of aggressiveness — infection efficiency, latency period and sporulation capacity

— for 44 isolates representative of the four *P. triticina* pathotype-genotype combinations. We

showed, for both pathotypes, that the most recent genotypes were more aggressive than the

older ones. Our findings were highly consistent for the various components of aggressiveness

for pathotype 166 317 0 grown on Michigan Amber — a 'naive' cultivar never grown in the

landscape — or on Apache — a 'neutral' cultivar, with no selection effect on the landscape-

pathotype pattern. For pathotype 106 314 0, the most recent genotype was more aggressive

on several of the cultivars most frequently grown in the landscape, but not on 'neutral' and

'naive' cultivars, and only in terms of its latency period. We conclude that the quantitative

components of aggressiveness can be significant drivers of evolution in pathogen populations.

A gain in aggressiveness allowed the maintenance of a declining pathotype, and even further

expansion of that pathotype, in the pathogen population, providing evidence that virulence

alone is not sufficient, aggressiveness also being required for the adaptation of a pathogen to

a changing varietal landscape.

**Keywords**

Aggressiveness, quantitative phenotyping, genotype evolution, host adaptation, leaf rust

**INTRODUCTION**

Plant diseases and pests cause crop damage accounting for up to 40% of yield losses

(Boonekamp, 2012). Pathogenicity, or the ability of plant pathogens, especially fungi, to cause

disease, is generally broken down into a qualitative term, 'virulence', and a quantitative term,

'aggressiveness' (Lannou, 2012). Virulence is defined as the capacity of the pathogen to infect

its host (compatible interaction), as opposed to avirulence, which expresses the resistance of

the host (incompatible interaction), according to the gene-for-gene model (Flor, 1971). A

virulence phenotype, also known as a pathotype or race, is defined by a virulence profile: two

pathogenic isolates are considered to belong to the same pathotype if they have the same

combination of virulence genes. Aggressiveness, the quantitative variation of pathogenicity

on a compatible host (Pariaud *et al.*, 2009a), can be viewed as the detrimental impact of a

pathogen on its host, leading to damage to the crop plant and, thus, yield losses (Shaner *et*

*al.*, 1992; Pariaud *et al.*, 2009a; Lannou, 2012). Aggressiveness also determines the rate at

which a given disease intensity is reached. Its assessment is intrinsically complex because it is

related to various life-history traits of the pathogen specific to its biology and the nature of

the symptoms that it produces. The different components of aggressiveness can be measured

by evaluating several of these complementary quantitative traits expressed during the host–

pathogen interaction. The most widely assessed aggressiveness components for rust

pathogens are infection efficiency, latency period and sporulation capacity (Pariaud *et al.*,

2009a; Lannou, 2012; Azzimonti *et al.*, 2013).

Infection efficiency is calculated by determining the proportion of spores able to cause a new

infection when deposited on compatible host plant tissues (Sache, 1997). The estimation of

this component is complex due to technical issues, particularly the need for great precision in

the spore deposition process (Lehman & Shaner, 1997), which involves placing a fixed and

known number of spores — ideally one by one — on the leaf.

The latency period is the length of time between the initiation of the infection process by a

unit of inoculum and the start of infectious unit production (Parlevliet, 1979; Madden *et al.*,

2007). In rusts, it is often defined as the length of time between inoculation and the

appearance of 50% of the sporulating structures (e.g. uredinia), also known as 'pustules' (Parlevliet, 1975; Johnson, 1980; Pfender, 2001). Latency period estimations therefore require counts of uredinia on at least a daily basis. The latency period is highly temperature-dependent, and its expression in thermal time is therefore recommended, to allow comparisons between trials (Lovell *et al.*, 2004).

Sporulation capacity is assessed as the number of spores produced per individual sporulating structure and per unit time (Sache, 1997; Pariaud *et al.*, 2009a). Spores can be collected and counted directly (e.g. with a cell counter) or indirectly (e.g. weighed) (Imhoff, 1982; Robert *et al.*, 2004; Delmotte *et al.*, 2014). However, in rusts, pustule density affects the number of spores produced (Robert *et al.*, 2004) and must, therefore, be taken into account in some analysis (Lannou & Soubeyrand, 2017). Moreover, sporulation is a continuous process, so sporulation capacity is time-dependent. This trait, like several others, is, thus, dependent on latency period. The interdependence of traits can be reduced by measuring sporulation capacity at a normalized time point.

Leaf rust caused by *Puccinia triticina* is one of the most damaging wheat diseases, causing high yield losses worldwide (Huerta-Espino *et al.*, 2011; Savary *et al.*, 2019). Qualitative resistance is the easiest and most effective means of limiting leaf rust epidemics. Eighty-two *Lr* genes have been identified in wheat cultivars, most displaying qualitative interactions (Bariana *et al.*, 2022). It has been shown that the deployment of qualitative resistance genes in the landscape exerts a strong selective pressure, acting as an essential driver of evolution in *P. triticina* populations (Goyeau *et al.*, 2006). This effect was highlighted by surveys of virulence phenotypes (pathotypes), which showed that the corresponding virulence can emerge rapidly

after the introduction of a new *Lr* gene into cultivars. For example, in France, virulence against

*Lr*28 appeared within only two years of the release of cultivars carrying *Lr28* (Fontyn *et al.*,

2022). Adaptation to qualitative resistances occurs rapidly, despite the clonality of the

population, through 'boom-and-bust' cycles of resistance (McDonald & Linde 2002).

However, qualitative resistance alone cannot fully explain pathogen population dynamics.

Indeed, the quantitative resistance genes introgressed into new wheat cultivars at the same

time as qualitative resistance genes may also exert a selection pressure in favor of more

aggressive pathotypes. The occurrence of selection for greater aggressiveness has already

been established for various pathogens (Delmas *et al.*, 2016; Frézal *et al.*, 2018). Milus *et al.*

(2009) showed that the replacement of an 'old' *Puccinia striiformis* f. sp. *tritici* population by

a 'new' population could be explained by the greater aggressiveness of isolates from this new

population. A survey of the French *P. triticina* population from 1999-2002 (Goyeau *et al.*, 2006)

revealed the domination of a single pathotype (073 100 0), coinciding with a period in which

the cultivar landscape was dominated by the cultivar Soissons. Pathotype 073 100 0 was found

to be more aggressive on this cultivar than other virulent pathotypes present in the *P. triticina*

population during this period (Pariaud *et al.*, 2009b). Fontyn *et al.* (2022) recently showed that

the domination of the French landscape by two pathotypes — 106 314 0 and 166 317 0 —

during the 2005-2016 period could not be fully attributed to the deployment of *Lr* genes.

Indeed, several other compatible pathotypes virulent against the *Lr* genes carried by the most

widely grown cultivars were present in the landscape, but never reached substantial

frequencies. The authors suggested that aggressiveness might drive the evolution of *P.*

*triticina* populations, modifying pathotype frequencies at large spatiotemporal scales. The

variation of aggressiveness over time, during the complete life history of a pathotype,

corresponding to the stages of emergence, domination and replacement, has never been

investigated.

The objective of this study was to determine the role of aggressiveness in the changes of *P.*

*triticina* pathotypes frequencies over large spatiotemporal scales during their life history. To

this end, we focused on two major pathotypes, 166 317 0 and 106 314 0, identified as good

experimental case studies for this purpose on the basis of investigations of frequency

dynamics over the 2005-2016 period in France. We first characterized isolates of these two

pathotypes genotypically with microsatellite markers, to identify potential 'subpopulations'.

Within each of the two pathotypes, we then compared the aggressiveness components of

isolates from the most frequent genotypes (i) on 'neutral' cultivars, i.e. cultivars with no

apparent effect on the frequencies of these pathotypes (166 317 0 and 106 314 0), and, when

relevant, (ii) on five cultivars widely grown in France over the study period (pathotype 106 314

0 only).

**MATERIALS AND METHODS**

**Selection and purification of isolates**

Annual surveys of *P. triticina* populations have been carried out at INRAE BIOGER over the last

two decades. These surveys involve the collection of leaf samples from field (micro)plots sown

with a single variety in a network of assays and nurseries throughout the wheat-growing areas

of France (Goyeau *et al.*, 2006; Fontyn *et al.*, 2022). In total, 2796 leaves were collected from

the 10 most cultivated varieties during the 2006-2016 period in the framework of the national

survey. Urediniospores were bulk-harvested from each leaf and stored at -80°C. One single-

pustule isolate was selected from each bulk and its pathotype was determined as described

by Goyeau *et al.* (2006). In total, 932 isolates were identified as pathotype 106 314 0 and 473

isolates were identified as pathotype 166 317 0. For the purposes of this study, we selected

286 urediniospore bulks of pathotype 106 314 0 and 115 from pathotype 166 317 0, as

identified during the national survey (Table S1 and S2). The bulks were defrosted and

repurified by the re-inoculation of seven-day-old cv. Michigan Amber wheat seedlings, to

obtain 401 new single-pustule isolates. Before inoculation, the plants were grown in cabinets

with air filters in a greenhouse at temperatures between 15 and 20°C, under a 14-h

photoperiod (daylight supplemented with 400 W sodium lamps). Plants were treated with 15

mL maleic hydrazide solution (0.25 g maleic hydrazide per liter of $H_2O$) to prevent the

emergence of secondary leaves and to increase spore production. Inoculated seedlings were

placed in a dew chamber at 15°C for 24 h and were then transferred to the greenhouse. One

week after inoculation, the seedlings were trimmed such that only one plant with one

uredinium remained in each pot. Before sporulation, cellophane bags were placed over the

pots to prevent contamination between isolates. Ten days after inoculation, 401 leaf

segments, each carrying only one uredinium, were collected for DNA extraction and

genotyping with microsatellite markers. Other single-pustule isolates were also selected from

28 and 16 bulks initially identified as pathotypes 106 314 0 and 166 317 0, respectively. Each

of these isolates was pathotyped (Goyeau *et al.*, 2006) and one of these isolates, after

confirmation that its virulence phenotype was as expected, was selected for DNA extraction

and genotyping with microsatellite markers. Spores from these 44 isolates were stored at -

80°C for further assessments of aggressiveness (Table S3). The various stages in the

purification, selection, pathotyping and genotyping of *P. triticina* isolates for which

aggressiveness components were evaluated are summarized in Figure 1.

**Genotyping of isolates with microsatellite markers**

<u>DNA extraction</u>

DNA was extracted from all the purified isolates in 96-well extraction plates, with Qiagen

DNeasy® Plant Mini Kit buffers (Qiagen, Hilden, Germany). To this end, each leaf segment

carrying a single uredinium was placed in a Qiagen collection microtube with a tungsten bead

and 100 µL of hot AP1 buffer (65°C). Leaf segments were ground by shaking the microtubes in

a Retsch® MM400 ball mill twice, for 30 seconds each, at 25 Hz. The tubes were then

centrifuged for 1 minute at 3000 x *g*. AP1 buffer supplemented with RNase A and Reagent DX

was then added to each tube (300 µL). After mixing, we added 130 µL P3 buffer to the tube,

which was then incubated for 10 minutes at -20°C and centrifuged at 4°C for 10 minutes at

20 000 x *g*. The supernatant (200 µL) was transferred to a new tube, to which we added 1

volume of sodium acetate (3 M pH 5) and 3 volumes of isopropanol (100%). After mixing, the

tubes were placed at -20°C for 30 minutes and then centrifuged for 20 minutes at 6000 x *g*.

The supernatant was removed and 3 volumes of 70% ethanol were added. The tube was

placed at -20°C for 5 minutes and was then centrifuged at 4°C for 15 minutes at 6000 x *g*. The

pellet containing the DNA was allowed to dry overnight, and was ten resuspended in 100 µL

ultra-purified water. We transferred 20 µL of the resulting suspension to Qiagen elution

microtubes RS in a 96-tube rack, which was sent to Eurofins (Eurofins, Luxembourg) for

genotyping. The DNA suspensions had concentrations of between 1 and 25 ng/µL.

<u>Microsatellite genotyping and analyses</u>

The 401 and 44 single-pustule isolates were genotyped for 19 microsatellite markers: RB8,

RB11, RB12, RB17, RB25, RB26, PtSSR13, PtSSR50, PtSSR55, PtSSR61, PtSSR68, PtSSR91,

PtSSR92, PtSSR152, PtSSR154, PtSSR158, PtSSR164, PtSSR173, and PtSSR186 (Duan *et al.*,

2003; Szabo & Kolmer, 2007). The microsatellite markers were assembled into two

multiplexes of 9 and 10 markers and labeled with four fluorochromes (Table S4) to prevent

overlaps between markers with the same range of allele sizes. PCR amplification was

performed by Eurofins (Eurofins, Luxembourg), with the following amplification program:

95°C for 5 min and 35 cycles of 95°C for 30 s, 58°C for 30 s, 72°C for 30 s, and then 60°C for 30

min. Each reaction contained 5 μL DNA solution and PCR mixture with Taq Type-it (Qiagen®).

PCR products were analyzed by capillarity electrophoresis on an ABl3130xl sequencer. The

two-multiplex PCR system was tested and validated on eight reference *P. triticina* DNA

extracts from isolates obtained from bread and durum wheat. Chromatograms were visually

inspected for all markers and for all individuals with Peak Scanner software version 2.0, before

the final assignment of SSR alleles. The chromosome position of SSR markers was determined

by performing a blastn analysis (with default parameter values) of the sequence of each

primer against the two genome sequences of the Australian isolate Pt76 (Duan *et al.*, 2021)

with the NCBI blast tool (https://blast.ncbi.nlm.nih.gov/Blast.cgi).

The genotyping of the 401 isolates revealed two major genotypes within each pathotype: 106

314 0-G1 and 106 314 0-G2 for pathotype 106 314 0, and 166 317 0-G1 and 166 317 0-G2 for

pathotype 166 317 0. These genotypes were taken into account in the design of subsequent

experiments.

**Evaluation of aggressiveness components**

<u>Experimental design</u>

Three components of aggressiveness were assessed for 28 isolates of pathotype 106 314 0

and 16 isolates of pathotype 166 317 0 that had been purified, pathotyped and genotyped

(Table S3): latency period, infection efficiency and sporulation capacity. These aggressiveness

components were measured in the greenhouse on seedlings of two wheat varieties: (i)

Apache, a commercial French cultivar characterized as 'neutral' in a previous study (Fontyn *et*

*al.*, 2022), i.e. with no selection effect on the landscape-pathotype pattern, and (ii) Michigan

Amber, considered intrinsically 'naive' or 'neutral', both because it carries no known leaf rust

resistance gene and because it has never been cultivated in France and cannot, therefore,

have played any significant role in the evolutionary trajectory of *P. triticina* in France.

It was not possible to perform a large single trial at our facilities, so the 44 isolates were

characterized in five successive series, according to the same protocol and under the same

experimental conditions. In series 1 and 2, we investigated the variation of aggressiveness in

pathotypes 106 314 0 and 166 317 0, distinguishing the two genotypes of each pathotype (106

314 0-G1 and 106 314 0-G2, 166 317 0-G1 and 166 317 0-G2, respectively) on Apache and

Michigan Amber. The isolates used for these two series were collected between 2005 and

2016 on Apache (Table 1). Series 3, 4 and 5 were performed to investigate the variation of

aggressiveness in 106 314 0 only. Again, the two genotypes (106 314 0-G1 and 106 314 0-G2)

were distinguished, this time on three additional cultivars from the following list — Aubusson,

Premio, Michigan Amber, Sankara, Expert and Bermude — all of which, except Michigan

Amber, were among the 35 most frequently grown cultivars in the French landscape during

the 2006-2016 period. The isolates used in these three series were collected in 2012 and 2013,

on Aubusson, Premio and Apache (Table 2). The isolates used in series 3 were collected on

cultivars Aubusson and Premio and were tested on these two cultivars and Michigan Amber.

The isolates used in series 4 were collected on cultivar Apache and were tested on Apache

and Michigan Amber. The isolates used in series 5 were collected on cultivar Apache and were

tested on cultivars Sankara, Expert and Bermude. In all five series, each isolate × cultivar

interaction was replicated eight times, on eight wheat seedlings. Three replicates of the

experimental design, including the five series, were established.

<u>Assessment of aggressiveness</u>

All the phenotyping tests were performed in a greenhouse on eight-day-old wheat seedlings

grown in a 21 x 15 x 6 cm plastic box containing potting soil placed under standardized

conditions (18°C at night and 22°C during the day, with a 16-h photoperiod). The second leaf

of each seedling was maintained, with double-sided tape, on a rigid plate coated with

aluminum foil and a 3 cm-long segment of this leaf was inoculated with the fungus. Inoculation

was performed with 10 fresh (two-week-old) spores, picked one by one with a human eyelash

under binocular magnifier and deposited on the leaf (Figure 2A and 2B). The plants were then

placed in a dew chamber, in the dark, at 15°C, for 18 to 24 h. They were then returned to the

greenhouse and the rigid plate was removed. When the first uredinia (pustules) began to

break through the leaf epidermis, generally six days after inoculation, they were counted at

10- to 14-hour intervals (twice daily) until no new pustules appeared (Figure 2C). Infection

efficiency (IE) was estimated for each leaf as the ratio between the final number of uredinia

and the number of spores deposited (10 spores). Latency period (LP), estimated as the time

between inoculation and the appearance of 50% of the total number of uredinia, was

expressed in degree-days, based on the air temperature measured in the greenhouse every

ten minutes. Sporulation capacity (SP) was estimated, once the number of uredinia had

stabilized, as the number of spores produced per uredinium over a four-day period. Once the

final number of uredinia had been reached (i.e. 9 days after inoculation) the spores that had

already been produced were removed from the leaf with a small brush. Slightly incurved

aluminum gutters (2 x 7cm) made from blind slats were then positioned under each inoculated

leaf, which was attached to the gutter with clips (Figure 2D). After four days, all the newly

produced spores (Figure 2E) were removed by suction with a cyclone collector into a portion

of plastic straw sealed at one end (Figure 2F). Each portion of straw was weighed before and

after spore harvesting. SP was calculated by dividing the total weight of the spores collected

from a single leaf by the number of uredinia on that leaf.

**Statistical analyses**

For each of the three aggressiveness traits (the response variable Y), two ANOVA models were

used: model (1) for series 1 and 2; model (2) for pooled data from series 3 and 4 (without a

series factor, as a preliminary analysis showed that the interactions of the series factor with

other factors were never significant) and for series 5, for each cultivar, as the cultivars tested

were not the same in the three series:

$$(1) \quad Y_{ijklm} = \mu + R_i + G_j + C_k + G_j/I_l + \varepsilon_{ijklm}$$

$$(2) \quad Y_{ijkl} = \mu + R_i + G_j + G_j/I_k + \varepsilon_{ijkl}$$

where $Y_{ijklm}$ and $Y_{ijkl}$ are the values of the studied trait in replicate (R) $i$, of genotype (G) $j$ and

isolate (I) $l$ or $k$, nested within genotype, on cultivar (C) $k$ (when this occurred). $\mu$ is the overall

mean value for this trait and $\varepsilon$ is the residual, representing the measurement error, with

$\varepsilon \sim N(o, \sigma2)$

Log, $\sqrt{}$ or 1/x transformation was applied to IE, LP and SP when necessary, to obtain a

normalized distribution of residuals. When the distribution of residuals could not be

normalized by any transformation, a non-parametric Kruskal-Wallis test was performed to

analyze the effect of genotype on the aggressiveness components.

All the analyses were performed with R software version 4.1.0.

**RESULTS**

**Frequency dynamics of two major *P. triticina* pathotypes in the French landscape in the**

**2006-2020 period**

The frequency of pathotype 106 314 0 in the French *P. triticina* population increased from 30%

in 2006 to 51% in 2009 (Figure 3). After a plateau at 30-33% from 2011 to 2014, the frequency

of this pathotype decreased strongly, to less than 1% in 2018. Within pathotype 106 314 0,

two main genotypes, G1 and G2, were identified and were considered to be dominant as their

cumulative frequency ranged from 40% to 65% during the 2006-2016 period (Figure 3). These

two genotypes differed at five of the 19 SSR loci studied (RB8, RB11, PtSSR68, PtSSR92 and

PtSSR164; Table 3). Genotype 106 314 0-G1 was the most frequent from 2006 to 2012, but it

decreased in frequency thereafter (Figure 4). Genotype 106 314 0-G2 was present at a very

low frequency from 2007 to 2011. In 2012, the frequencies of the two genotypes were fairly

similar, at 23% for 106 314 0-G1 and 30% for 106 314 0-G2. Genotype 106 314 0-G1 then

continued to decrease in frequency, eventually disappearing completely from the sampled

population in 2014. Conversely, genotype 106 314 0-G2 continued to increase in frequency. It

accounted for 42% of the total 106 314 0 pathotype population in 2013 and 66% in 2016. Over

the entire period, genotypes other than G1 and G2 were also identified (detailed data not

shown) within pathotype 106 314 0. Their cumulative frequency ranged from 40% to 60%

from 2006 to 2016.

Pathotype 166 317 0 first appeared in 2007, initially at a very low frequency (less than 2%). Its

temporal evolution in the French *P. triticina* population was characterized by two peaks, at

32% in 2012 and 41% in 2018, with a decrease in frequency to 13% in 2015 (Figure 3). The

frequency of this pathotype decreased after 2018 reaching 27% in 2020. Within pathotype

166 317 0, two genotypes, differing by only one of the 19 SSR loci used (RB8; Table 3), were

identified and considered to be predominant during the 2013-2016 period. In 2013 and 2014,

genotype 166 317 0-G1 dominated, as it accounted for more than 65% of the total 166 317 0

pathotype population (Figure 5). In 2015, the 166 317 0-G2 genotype emerged and became

co-dominant with 166 317 0-G1. In 2016, 166 317 0-G2 sharply increased in frequency, to 62%,

whereas the frequency of 166 317 0-G1 decreased to 7%. Over the four-year period

considered, several other genotypes were identified (detailed data not shown). Their

cumulative frequencies reached 46% during the co-dominance of the two major genotypes,

and 30% when a single genotype dominated.

**Differences in aggressiveness between genotypes of the two major *P. triticina* pathotypes**

**expressed on 'neutral' and 'naive' cultivars**

Genotype 166 317 0-G2 was more aggressive than genotype 166 317 0-G1 in assessments of

all three aggressiveness components on a 'neutral' cultivar, Apache, and a 'naive' cultivar

Michigan Amber (in analyses of the pooled dataset for the two cultivars; Figure 6A, B and C).

The data for assessments of aggressiveness components on both cultivars are provided in the

supplementary data (Figure S1). Infection efficiency (IE) was higher for genotype 166 317 0-

G2 (53.1 %) than for genotype 166 317 0-G1 (47.6%). Sporulation capacity (SP) was also higher

for genotype 166 317 0-G2, with 0.120 mg/lesion versus 0.114 mg/lesion for genotype 166

317 0-G1. Latency period (LP) differed significantly between the two genotypes; it was shorter

for genotype 166 317 0-G2 (132.6 vs 135.0 degree-days). Significant differences were found

between isolates of the two genotypes.

Infection efficiency (IE), latency period (LP) and sporulation capacity (SP) did not differ

between genotypes 106 314 0-G1 and 106 314 0-G2 assessed on cultivars Apache and

Michigan Amber (in analyses of the pooled dataset for the two cultivars; Figure 6D, E and F).

The data for assessments of aggressiveness components on both cultivars are provided in the

supplementary data (Figure S2). The IE for genotype 106 314 0-G1 was 43.8%, which is not

significantly different from the value of 43.4% obtained for genotype 106 314 0-G2. SP was

0.128 mg/lesion for genotype 106 314 0-G1, which is not significantly different from the value

of 0.124 mg/lesion obtained for genotype G2. Finally, LP differed by less than one degree-day

between the two genotypes, at 133.1 degree-days for genotype 106 314 0-G1 and 133.3

degree-days for genotype 106 314 0-G2. By contrast to our findings for pathotype 166 317 0,

we observed no significant difference between genotypes of the same pathotype for 106 314

0. This singularity justified our decision to perform phenotypic characterization by comparing

the aggressiveness of isolates of pathotype 106 314 0 on other cultivar sets with a view to

detecting potential differences.

**Difference in aggressiveness between the two major genotypes of pathotype 106 314 0 on**

**Michigan Amber and major cultivars present in the French varietal landscape**

The two major genotypes of pathotype 106 314 0 had significantly different latency periods

(LP) on five of the six cultivars tested (Table 4). These differences were significant on the

cultivar Michigan Amber, whereas no difference was found in analyses of the data for series

1, as described above. The infection efficiency (IE) of genotype 106 314 0-G2 was higher than

that of 106 314 0-G1 (48.7% versus 39.1%) on cultivar Expert, but no differences were found

on the other cultivars. Genotype 106 314 0-G2 had a shorter LP than 106 314 0-G1 on five of

the six cultivars tested. This difference in LP ranged from 2.1 degree-days on Premio to 4.8

degree-days on Expert. A difference of 5 degree-days is equivalent to 8 h at 15°C. Bermude

was the only cultivar on which LP did not differ significantly between the two pathogen

genotypes. No significant differences in sporulation capacity (SP) were observed between the

two genotypes on any of the cultivars tested. SP varied between cultivars, ranging from 0.099

mg on Michigan Amber to 0.134 mg on Sankara, with slightly higher values for 106 314 0-G2

than for 106 314 0-G1. It was not possible to detect significant differences between cultivars

because different cultivars were analyzed in different series.

Within genotypes, there was a significant isolate effect for LP, but no significant isolate effect

for IE or SP, on Aubusson, Premio and Expert.

For LP, IE and SP, the replication factor (x3) was almost always significant. The interaction

between replication factor and the three aggressiveness components was significant for some

aggressiveness component measurements (Table 4).

**DISCUSSION**

Focusing on two major pathotypes from the French *P. triticina* population, we found that

several genotypes were present within each pathotype, and that the frequency of the most

common genotypes changed over time. The initially dominant genotypes representative of

pathotypes 166 317 0 and 106 314 0 were replaced, in each pathotype, by another genotype

over the period 2006-2016. The most recent dominant genotype was more aggressive than

the older one in both pathotypes.

**Methodological aspects of aggressiveness measurement at the wheat seedling stage**

The mean values used to characterize the aggressiveness of different *P. triticina* isolates on

wheat seedlings revealed significant variation for each of the three aggressiveness

components measured — infection efficiency, latency period and sporulation capacity — at

the seedling stage. Aggressiveness and its variation can be measured on seedlings (Milus *et*

*al.*, 2006; de Vallavieille-Pope *et al.*, 2018) or on adult plants (Lehman & Shaner, 1997; Pariaud

*et al.*, 2009b; Azzimonti *et al.*, 2013). Plant stage is known to affect the aggressiveness

components of *P. triticina* (Pariaud *et al.*, 2009b). We decided to assess aggressiveness

components on wheat seedlings in this study, because the use of seedlings in a semi-

controlled environment results in more homogeneous physiological properties of the plant at

inoculation and because this approach requires less space and time than studies on adult

plants.

With the phenotyping method developed and used in this study, we were able to estimate the

three aggressiveness components simultaneously on the same inoculated plant. We

measured infection efficiency (IE) more precisely here than in previous studies, in which the

variability was sometimes very high. For instance, Pariaud *et al.* (2009b) applied a 1:10 mixture

of *P. triticina* urediniospores and *Lycopodium* spores to the leaf surface with a soft brush; they

reported IE values ranging from 18% to 80% for the same genotype. This high level of variation

resulted from difficulty controlling the number of urediniospores deposited with this method,

resulting in differences between experimental series. Although fastidious, the deposition of

urediniospores one-by-one on the leaf surface resulted in a much more accurate estimation

of IE than other methods based on the dilution of *P. triticina* urediniospores among

*Lycopodium* spores or in liquids, such as mineral oils (Pariaud *et al.*, 2009; Sørensen *et al.*,

2016). Nevertheless, the interaction between IE and the replication factor was almost always

significant, indicating an impact of environmental conditions on this component of

aggressiveness. This interaction reflects the difficulty ensuring uniform dew quality in the dew

chamber just after inoculation, as the success of urediniospore germination and penetration

depends on the presence of a water film on the leaf surface (Bolton *et al.*, 2008).

Sporulation capacity (SP) was measured by collecting urediniospores produced between 9 and

12 days after inoculation, and differences in SP between genotypes were detected for only

one experimental series. Pariaud *et al.* (2009b) found differences in SP between pathotypes

of *P. triticina* in a study in which urediniospores were collected 15 to 23 days after the

inoculation of adult plants. They also found that the difference in SP between isolates

increased with the regular collection of urediniospores until 59 days after inoculation. The

collection of urediniospores more than 13 days after inoculation is of potential relevance for

future experiments with our method, to maximize differences and reveal small differences in

this component of aggressiveness between isolates.

Latency period (LP) was the component that differed most between pathogen genotypes and

was the least affected by the replication factor. The calculation of LP in degree-days made it

possible to take the temperature-dependence of both infection processes and pathogen

development in the leaf into account (Lovell *et al.*, 2004). In most pathosystems, latency

period is a component of choice for studies of quantitative interactions (aggressiveness,

partial resistance) because time is a variable that is much easier to fractionate and, therefore,

to quantify precisely, than any biological trait, making it possible to highlight extremely small

differences between isolates repeatedly. This is almost an epistemological issue for

experimental epidemiology and phytopathometry (Suffert & Thompson, 2018; Bock *et al.*,

2022).

**Evolution of greater aggressiveness in two *P. triticina* pathotypes**

In the case of pathotype 166 317 0, genotype G2, which replaced genotype G1, was more

aggressive than its predecessor, with highly consistent results obtained for the different

components of aggressiveness. Following this switch between genotypes, the frequency of

pathotype 166 317 0 increased considerably over a period of three years, whereas the

proportion of compatible host cultivars in the landscape decreased (data not shown). Severe

*P. triticina* epidemics have been shown to be associated with a high sporulation capacity, a

high infection efficiency and a short latency period (Azzimonti *et al.*, 2022). In pathotype 106

314 0, comparisons of aggressiveness on a 'neutral' cultivar revealed no significant difference

between the most recent and oldest genotypes. A comparison of the aggressiveness of these

two genotypes on some of the cultivars mostly frequently grown in the landscape revealed

that 106 314 0-G2 was more aggressive than 106 314 0-G1 only in terms of its latency period.

This change in predominant genotype coincided with a halt in the decline of pathotype 106

314 0 frequency in the landscape. The shorter latency period of genotype 106 314 0-G2 seems

to have an effect on the frequency of this pathotype in the landscape, consistent with the

assertions of several studies that latency period is the aggressiveness component with the

largest effect on pathogen dynamics in field conditions (Lannou, 2012). In modeling studies,

this also appeared to be the trait with the largest impact on the intensity of *P. triticina*

epidemics, as it determined the number of reproductive cycles (i.e. from inoculation to the

spore dispersion) of the pathogen possible in a single season (Rimbaud *et al.*, 2018). Although

only latency period differed significantly between 106 314 0-G2 and 106 314 0-G1, the data

for the other aggressiveness components also supported the notion that 106 314 0-G2 was

more aggressive, for every component, on all cultivars, except for infection efficiency on

Premio. Measurements of aggressiveness components on seedlings in controlled conditions

(only one reproductive cycle of the pathogen) have a limited capacity for the detection of

small phenotypic variations.

The competitive advantage of one particular genotype within a pathotype, expressing a

difference in fitness for the same virulence profile, may account for the replacement of one

genotype with another. This hypothesis is particularly realistic here as the emergence of the

more aggressive genotype in each pathotype, 106 314 0 and 166 317 0, coincided with a short-

term change in the mid-term trend in frequency changes: (i) the decrease in the frequency of

pathotype 106 314 0 in the landscape observed from 2009 to 2011 was temporarily halted

during the 2011-2014 period, when the frequency of this pathotype reached a plateau (Figure

2), coinciding with the replacement of the less aggressive genotype 106 314 0-G1 with the

more aggressive genotype 106 314 0-G2 in 2012 and 2013 (Figure 3); (ii) the decrease in the

frequency of pathotype 166 317 0 in the landscape observed from 2012 to 2015 was halted

over the 2015-2018 period, resulting in a new peak frequency (Figure 2) coinciding with the

replacement of the less aggressive 166 317 0-G1 genotype with the more aggressive 166 317

0-G2 genotype in 2012 and 2013 (Figure 3).

These results suggest that the replacement of some pathotypes by others, driven by changes

in the frequencies of resistance genes in the varietal landscape, may be slowed, to various

extents, by increases in the aggressiveness of certain genotypes or 'lineages' (defined as

genetically related genotypes of the same group of pathotypes, i.e. with the same close

ancestor), consistent with the recent analysis by Fontyn *et al.* (2022). It is even possible, as for

pathotype 166 317 0, for the pathotype to gain a 'second life' — for a period of three to four

years — due to an increase in the aggressiveness of a new genotype, as long as it remains

adapted (or at least not maladapted) to the resistance genes present in the varieties deployed

in the landscape. This empirical study shows that differences in aggressiveness can be

expressed on a 'neutral' cultivar (for pathotype 166 317 0), but sometimes (for pathotype 106

314 0) only on 'non-neutral' cultivars representative of a varietal landscape. This finding

clearly complicates the experimental approach (difficulty determining the most appropriate

experimental design without making assumptions), but also the data analysis. The magnitude

of the differences in aggressiveness probably depends on the type of cultivars tested ('neutral'

or 'non-neutral'). Indeed, we found that the intensity of the mid-term trend in the change in

frequencies highlighted above was variable, with either a simple slowing of a trend towards a

decrease (plateau, as for pathotype 106 314 0), or a change in direction, with a new increase

(peak, as for pathotype 166 317 0).

**Greater aggressiveness as a selective advantage within plant pathogen populations**

Increases in aggressiveness have already been proposed as an explanation for the shifts

observed in rust populations. Milus *et al.* (2009) measured five aggressiveness components

for *P. striiformis* f. sp. *tritici* isolates with similar virulence profiles collected in the U.S. before

or after 2000. Isolate aggressiveness was assessed on a susceptible cultivar, and comparisons

revealed that the most recent isolates were more aggressive than the older ones (Milus *et al.*,

2009). Studies on host quantitative resistance effects on pathogen populations have revealed

directional selection toward an increase in pathogen aggressiveness on host cultivars

(Andrivon *et al.*, 2007; Frézal *et al.*, 2018). Isolates of *Phytophthora infestans* collected on a

partially resistant potato cultivar at the end of the epidemic produced larger lesions and had

higher sporulation rates than those collected at the beginning of the epidemic (Andrivon *et*

*al.*, 2007). *Z. tritici* isolates collected from a partially resistant wheat cultivar were found to be

more aggressive than isolates collected from a susceptible wheat cultivar (Cowger & Mundt,

2002). Isolates of *Venturia inaequalis* collected from an apple cultivar carrying a QTL for

resistance had higher lesion densities and sporulation rates than isolates collected from a

susceptible variety (Caffier *et al.*, 2016). Isolates of *Plasmopara viticola* collected from partially

resistant varieties were found to have shorter latency periods and higher rates of spore

production than isolates collected from susceptible cultivars (Delmas *et al.*, 2016). Many

studies have shown that the use of quantitative sources of resistance affects the evolution of

pathogen population by leading to the selection of more aggressive isolates. Quantitative

interactions based on partial resistance in the host and aggressiveness in the pathogen are a

significant driver of evolution in pathogen populations. We show here that increases in

aggressiveness were associated with genotype replacement in two major *P. triticina*

pathotypes, consistent with all the examples of selection based on quantitative plant-

pathogen interactions cited above.

**Putative origin of the more aggressive genotypes**

All surveys in wheat-growing areas to date have provided evidence for continual evolution of

*P. triticina* populations, with rapid changes in pathotype frequencies (Goyeau *et al.*, 2012;

Kosman *et al.*, 2019; Zhang *et al.*, 2020; Fontyn *et al.*, 2022; Kolmer & Fajolu, 2022). These

changes can be explained intrinsically by the acquisition of new virulence genes, by mutations,

somatic exchanges or, more rarely, genetic recombination. A study on the Australian *P.*

*striiformis* f. sp. *tritici* population showed that it had evolved due to mutation events (Steele

*et al.*, 2001). Whole-genome sequencing has detected similar recurrent mutations in *P.*

*triticina*, suggesting that such events play a major role in genetic variability within clonal

lineages (Fellers *et al.*, 2021). Somatic exchange may also have played a role, as in the

emergence of the Ug99 lineage of *Puccinia graminis* f. sp. *tritici,* which had one identical

haploid nucleus in common with an old African lineage, with no sign of recombination (Li *et*

*al.*, 2019). A similar situation has been described in *P. triticina*, with the Australian isolate Pt64

resulting from somatic exchange between two parental isolates (Wu *et al.*, 2019). More rarely,

genetic recombination may also underlie the emergence of new genotypes with new virulence

genes. Indeed, *P. triticina* is a heteroecious fungus that requires an alternative host,

*Thalictrum speciosissimum* (not naturally present in most places worldwide, including France),

for sexual reproduction. The high proportion of repeated genotypes and the heterozygosity

rates of European and French leaf rust populations confirmed the very dominant role of clonal

reproduction (Goyeau *et al.*, 2007; Kolmer *et al.*, 2013). Genetic changes do not necessarily

result in different virulence profiles, so isolates with identical or very similar pathotypes may

have different origins. In a previous study on *P. triticina* nine microsatellite markers were used

to analyze the genotypes of 33 pathotypes (Goyeau *et al.*, 2007). This analysis highlighted the

presence of different genotypes, differing by only one of the nine SSRs tested, in five of the

pathotypes. Another study on 121 European *P. triticina* isolates with 21 SSR markers revealed

a significant correlation between phenotype and genotype, but the same phenotype could,

nevertheless, be associated with several genotypes (Kolmer *et al.*, 2013). The results of our

study raise questions as to the genetic relationship between the most recent genotypes, 166

317 0-G2 and 106 314 0-G2, and the oldest genotypes, 166 317 0-G1 and 106 314 0-G1,

respectively. Can the most recent and the oldest genotypes be considered to belong to the

same 'lineage', as previously defined? A mutation event during asexual reproduction probably

led to the emergence of the most recent genotype (G2) of pathotype 166 317 0, which differs

from the older genotype (G1) by only one SSR allele (Table 3). By contrast, genotypes G1 and

G2 of pathotype 106 314 0 displayed a higher degree of genetic diversity, differing by five SSR

alleles (Table 3). Genotype 106 314 0-G2 may, therefore, be the result of several mutation

events occurring in the same lineage and conferring greater aggressiveness than for 106 314

0-G1. Alternatively, the presence of 106 314 0-G2 in France may reflect the introduction from

an exotic source of a different (more aggressive) lineage that had acquired the same virulence

genes and, therefore, belonged to the same pathotype, 106 314 0. Migration events are

frequent in cereal rust populations, as already shown for yellow rust (Ali *et al.*, 2014; Bueno-

Sancho *et al.*, 2017), for which there is a famous example of migration in the form of an

introduction into Australia from northwest Europe in 1979 (Wellings & McIntosh, 1990).

Studies on *P. triticina* in wheat-growing areas worldwide have revealed a broad geographic

distribution of identical and highly related multilocus genotypes, highlighting the potential of

leaf rust for long-distance migration (Kolmer *et al.*, 2019) and the possibility of new genotypes

resulting from exotic introduction.

**Additional approaches for detecting genetic variation related to aggressiveness**

Pathotypes of *P. triticina*, defined on the basis of their virulence profiles, appear to be

represented by several genotypes, the distribution of which changes over time. An analysis of

genotypes within the two major pathotypes collected in France in 2006-2020, 166 317 0 and

106 314 0, revealed the presence in each pathotype of two main genotypes, one of which

replaced the other over time. This study was not designed to characterize overall pathotypic

or genotypic changes in the *P. triticina* population, so our data should be extrapolated with

caution, as any attempt to use them for such purposes would be subject to multiple sampling

biases. In addition to the four genotypes on which we focused here, several other genotypes

(with cumulative frequencies of 31% to 61%, depending on the year) were identified. This is

due to the initial pathotyping performed annually for the national survey and the genotyping

for this study being performed on different isolates, purified from a bulk of urediniospores

collected from a single leaf, which may well have been infected with several pathotypes

and/or genotypes. This highlights the imperfections of the protocol linked to the constraints

of working on datasets and biological material acquired over several decades with ever-

changing techniques. It would have been better to pathotype and phenotype the 401 isolates

from the same purified batch of urediniospores, as we did later on for the 44 isolates for which

aggressiveness was characterized (Figure 1). However, given the size of the sample, this

limitation did not prevent us from obtaining a correct overview of changes for the most

common genotypes.

Our results revealed rapid genotype evolution within the pathotypes that was not detectable

if only virulence phenotypes were considered. These findings demonstrate the value of

analyzing population dynamics not only from the pathotype standpoint, but also with

genotype data, to obtain a more informative picture of pathogen diversity. The use of a small

number of SSR markers is not sufficient for the detection of genetic variations in *P. triticina*

populations. Genome-wide genotyping approaches are required to characterize the genetic

diversity of leaf rust populations more precisely and to explain the emergence of new

genotypes. As an illustration of this approach, a combined genome and transcriptome analysis has been performed on 133 *P. striiformis* f. sp. *tritici* isolates collected in 16 European countries. This analysis provided more precise information about the origin of the new emerging races, and showed that SNP analysis was an effective approach for the detection of pathogen diversity and for pathogen surveillance (Bueno-Sancho *et al.*, 2017). Similarly, Fellers *et al.* (2021) genotyped 121 *P. triticina* isolates with 121 907 SNP markers, and showed that recurrent mutation and selection had played a major role in differentiation within clonal lineages. The results of our study highlight the importance of combining genome-wide genotyping tools with precise pathotyping and aggressiveness phenotyping to detect the emergence of new variants and to improve our understanding of population dynamics.

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**Data Availability Statement**

The data that support the findings of this study are openly available in the INRAE Dataverse

online data repository (https://data. inrae. fr/) at https://doi.org/10.57745/MZ8TDK

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**TABLES**

**Table 1.** Experimental design, with the allocation of isolates to series 1 and 2, for assessments of the aggressiveness of pathotypes 106 314 0 and 166 317 0. Isolates were collected during the 2005-2016 period, from cultivar Apache, and were tested on Apache and Michigan Amber.

| Pathotype | Genotype[a] | Year of sampling | Number of isolates | Series |
|---|---|---|---|---|
| 106 314 0 | 106 314 0-G1 | 2005-2006 | 6 | 1 |
| | | 2012-2013 | 3 | 1 |
| | | 2015-2016 | 0 | 1 |
| | 106 314 0-G2 | 2005-2006 | 0 | 1 |
| | | 2012-2013 | 2 | 1 |
| | | 2015-2016 | 6 | 1 |
| 166 317 0 | 166 317 0-G1 | 2012 | 6 | 2 |
| | | 2014 | 6 | 2 |
| | | 2016 | 0 | 2 |
| | 166 317 0-G2 | 2012 | 0 | 2 |
| | | 2014 | 0 | 2 |
| | | 2016 | 4 | 2 |

[a]genotyping wirh19 SSRs specific to leaf rust (see Table 3)

**Table 2.** Experimental design, with the allocation of isolates to series 3, 4 and 5, for assessments of the aggressiveness of pathotype 106 314 0. Isolates were collected in 2012-2013, from Aubusson, Premio or Apache.

| Genotype[a] | Number of isolates | Cultivar tested | Series |
|---|---|---|---|
| 106 314 0-G1 | 5 | Aubusson | 3+4 |
| | | Premio | 3+4 |
| | | Michigan Amber | 3+4 |
| | 1 | Sankara | 5 |
| | | Expert | 5 |
| | | Bermude | 5 |
| 106 314 0-G2 | 11 | Aubusson | 3+4 |
| | | Premio | 3+4 |
| | | Michigan Amber | 3+4 |
| | 2 | Sankara | 5 |
| | | Expert | 5 |
| | | Bermude | 5 |

[a]genotyping with 19 SSRs specific to leaf rust (see Table 3)

**Table 3.** Genotypic characteristics of the two main genotypes, G1 and G2, identified for isolates of pathotypes 166 317 0 and 106 314 0, collected from 2005-2016 for 19 SSR loci. Allele sizes differing between genotypes of the same pathotype are shown in bold typeface.

| | 166 317 0[a] | | | | 106 314 0[a] | | | |
| | 166 317 0-G1 | | 166 317 0-G2 | | 106 314 0-G1 | | 106 314 0-G2 | |
| Locus | Allele 1 | Allele 2 | Allele 1 | Allele 2 | Allele 1 | Allele 2 | Allele 1 | Allele 2 |
|---|---|---|---|---|---|---|---|---|
| RB8 | **143** | **146** | **146** | **152** | **143** | **143** | **143** | **152** |
| RB11 | 200 | 200 | 200 | 200 | **200** | **204** | **175** | **204** |
| RB12 | 282 | 290 | 282 | 290 | 282 | 290 | 282 | 290 |
| RB17 | 190 | 190 | 190 | 190 | 190 | 190 | 190 | 190 |
| RB25 | 230 | 230 | 230 | 230 | 230 | 230 | 230 | 230 |
| RB26 | 352 | 352 | 352 | 352 | 352 | 352 | 352 | 352 |
| PtSSR13 | 129 | 131 | 129 | 131 | 129 | 131 | 129 | 131 |
| PtSSR50 | 365 | 371 | 365 | 371 | 365 | 371 | 365 | 371 |
| PtSSR55 | 308 | 308 | 308 | 308 | 308 | 308 | 308 | 308 |
| PtSSR61 | 296 | 302 | 296 | 302 | 296 | 300 | 296 | 300 |
| PtSSR68 | 309 | 311 | 309 | 311 | **309** | **317** | **309** | **323** |
| PtSSR91 | 380 | 382 | 380 | 382 | 382 | 382 | 382 | 382 |
| PtSSR92 | 246 | 246 | 246 | 246 | **246** | **246** | **246** | **248** |
| PtSSR152 | 389 | 393 | 389 | 393 | 389 | 393 | 389 | 393 |
| PtSSR154 | 247 | 267 | 247 | 267 | 247 | 267 | 247 | 267 |
| PtSSR158 | 235 | 238 | 235 | 238 | 232 | 238 | 232 | 238 |
| PtSSR164 | 219 | 219 | 219 | 219 | **219** | **225** | **219** | **219** |
| PtSSR173 | 213 | 221 | 213 | 221 | 213 | 221 | 213 | 221 |
| PtSSR186 | 341 | 341 | 341 | 341 | 341 | 341 | 341 | 341 |

[a] Seven-digit triplet code (Gilmour, 1973) based on a 20-*Lr* gene differential set:  Thatcher lines [*Lr1*, *Lr2a*, *Lr2b*], [*Lr2c*, *Lr3*, *Lr3bg]*, [*Lr3ka*, *Lr10*, *Lr13*], [*Lr14a*, *Lr15*, *Lr16*], [*Lr17*, *Lr20*, *Lr23*], [*Lr26*, *Lr17b* (the Australian cv. 'Harrier')*, Lr37*] and [*Lr24, , Lr28* (*line* CS2A/2M)].

**Table 4.** Comparison of the aggressiveness of the two major genotypes of pathotype 106 314 0 on five of the mostly widely grown cultivars in the French landscape during the 2005-2016 period. Letters indicate a significant difference between genotypes on the same cultivar after a Tukey multiple-comparison test or a Wilcoxon-Mann-Whitney test. Numbers in bold typeface indicate a significant difference between isolates within genotypes.

| Cultivar | Genotype | Aggressiveness components | | |
| --- | --- | --- | --- | --- |
| | | Infection efficiency$^z$ | Latency period$^z$ | Sporulation capacity$^z$ |
| Aubusson | 106 314 0-G1 | 48.5* | **141.0 a** | 0.105 |
| | 106 314 0-G2 | 49.8* | **137.3 b** | 0.106 |
| Premio | 106 314 0-G1 | 52.6* | **134.5 a** | 0.102 |
| | 106 314 0-G2 | 49.8* | **132.4 b** | 0.105 |
| Bermude | 106 314 0-G1 | 43.9* | 143.4 | 0.114* |
| | 106 314 0-G2 | 47.0* | 142.1 | 0.121* |
| Expert | 106 314 0-G1 | 39.1 a | **143.0 a** | 0.120* |
| | 106 314 0-G2 | 48.7 b | **138.2 b** | 0.132* |
| Sankara | 106 314 0-G1 | 45.0* | 143.9 a* | 0.134* |
| | 106 314 0-G2 | 46.9* | 140.7 b* | 0.134* |
| Michigan Amber | 106 314 0-G1 | 51.6* | 131.7 a | 0.099 |
| | 106 314 0-G2 | 52.2* | 129.3 b | 0.102 |

$^z$ Infection efficiency (IE) was measured as a %, latency period (LP) in degree-days, and sporulation capacity (SP) in mg of spores/uredinia.

* indicates a significant interaction between genotype and replication factors.

**FIGURES**

**Figure 1.** Overview of the purification, selection, pathotyping, and genotyping steps for the *P. triticina* isolates for which aggressiveness components were evaluated.

**Figure 2.** Experimental steps for the assessment of aggressiveness components of *Puccinia triticina* on wheat seedlings. Inoculation: (A) and (B), collection of a spore with a human eyelash and its deposition on a leaf. Latency period (C): (C1) onset of chlorosis, (C2) counting of the uredinia (pustules) that have broken through the leaf, (C3) end of the latency period, when all the pustules have emerged. Sporulation: (D) incurved aluminum gutters positioned under the leaves for spore collection, (E) end of sporulation, (F) spores retrieved with a cyclone collector into a sealed portion of plastic straw.

**Figure 3.** Frequency dynamics of two of the major *Puccinia triticina* pathotypes, 106 314 0 and 166 317 0, in the French landscape during the 2006-2020 period. Pathotype frequency was determined with data from the national survey (see Fontyn *et al.*, 2022).

**Figure 4.** Changes in genotype frequencies within pathotype 106 314 0 in the French landscape during the 2006-2016 period, as determined with 19 SSR markers. The numbers on top of the bars are the numbers of isolates genotyped.

**Figure 5.** Changes in genotype frequencies within pathotype 166 317 0 in the French landscape during the 2013-2016 period, as determined with 19 SSR markers. Numbers on top of the bars represent the numbers of isolates genotyped.

**Figure 6.** Comparison of genotype aggressiveness within the two major *Puccinia triticina* pathotypes. Infection efficiency (IE) as a %, latency period (LP) in degree-days and sporulation capacity (SP) in mg for pathotypes 166 317 0 (A, B, C) and 106 314 0 (D, E, F) were assessed on cultivars Apache and Michigan (pooled data, see Figures S1 and S2 for results by cultivar). Isolates were sampled from Apache in 2012, 2014, and 2016 for pathotype 166 317 0, and in 2005-2006, 2012-2013 and 2015-2016 for pathotype 106 314 0. Within a box plot, the black diamond represents the mean value and the bar indicates the median value. Letters indicate significant differences between genotypes in Kruskal-Wallis tests (A, C, E, F) or ANOVA (B and D).