

1 **Means, motive, and opportunity for biological invasions: genetic**
2 **introgression in a fungal pathogen**

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22 **ABSTRACT**

23 Invasions by fungal plant pathogens pose a significant threat to the health of agriculture
24 ecosystems. Despite limited standing genetic variation, many invasive fungal species can
25 adapt and spread rapidly, resulting in significant losses in crop yields. Here, we report on
26 the population genomics of *Colletotrichum truncatum*, a polyphagous pathogen that can
27 infect more than 460 plant species, and an invasive pathogen on soybean in Brazil. We
28 study the whole-genome sequences of 18 isolates representing 10 fields from two major
29 regions of soybean production. We show that Brazilian *C. truncatum* is subdivided into
30 three phylogenetically distinct lineages that exchange genetic variation through
31 hybridization. Introgression affects 2 to 30% of the nucleotides of genomes and varies
32 widely between the lineages. We find that introgressed regions comprise secreted

33 protein-encoding genes, suggesting possible co-evolutionary targets for selection in those
34 regions. We highlight the inherent vulnerability of genetically uniform crops in the agro-
35 ecological environment, particularly when faced with pathogens that can take full
36 advantage of the opportunities offered by an increasingly globalized world. Finally, we
37 discuss “The Means, Motive, and Opportunity” of fungal pathogens and how they can
38 become invasive species of crops. We call for more population genomic studies because
39 such analyses can help identify geographic areas and pathogens that pose a risk, thereby
40 helping to inform control strategies to better protect crops in the future.

41

42 **INTRODUCTION**

43 Understanding the eco-evolutionary and human-associated factors underlying
44 the emergence and spread of fungal plant diseases is essential to the
45 implementation of effective control measures (Hessenauer et al., 2020;
46 Stukenbrock & McDonald, 2008). Population genetics and molecular
47 epidemiology can shed light on both the extrinsic and intrinsic drivers of biological
48 invasions by fungal plant pathogens (Gladieux et al., 2015; Grünwald, McDonald,
49 & Milgroom, 2016). In parallel, molecular ecology and evolution can provide
50 insights into how an invasive species with limited genetic variation can evolve into
51 ecologically successful pest species soon after a founder event.

52 Fungi can rapidly evolve into devastating invasive species, despite their often
53 low genetic diversity (Ali et al., 2014; De Jonge et al., 2013; Gladieux et al., 2018;
54 Latorre et al., 2020; Stauber, Badet, Prospero, & Croll, 2020). Although this is a
55 characteristic that they share with many other invasive species, some unique
56 aspects of fungal population biology may facilitate rapid evolutionary changes and
57 enhance their invasive potential. Despite regular demographic bottlenecks (e.g.,

58 during winters), fungal populations generally have periods of huge census size,
59 which can substantial evolutionary changes (Gladieux et al., 2015).

60 Fungal reproduction is complex and can include both sexual and asexual stages
61 (Alexopoulos, Mims, & Blackwell, 1996). Asexual reproduction can be
62 accomplished through mitotic spores (conidia), which augments the propagule
63 pressure during invasion in non-native habitats. **In addition, fungi produce**
64 **filamentous structures, i.e., hyphae, which provide a second mode of asexual**
65 **reproduction through mycelial fragmentation.** Complementing these asexual
66 modes, during sexual reproduction, compatible individuals may exchange genetic
67 information during plasmogamy and karyogamy between gametes (Alexopoulos et
68 al., 1996). Meiotic spores promote the creation of novel genotypes through
69 recombination, and they serve as dispersal and survival structures (Taylor,
70 Jacobson, & Fisher, 1999). If compatible gametes are derived from genetically
71 diverged lineages, the resulting genetic exchange can lead to genetic introgression.

72 The large census size of invasive fungal populations enables rapid adaptations
73 to new varieties of resistant plants or antifungals molecules (Barton, 2010;
74 Gladieux et al., 2015). This may be particularly important for fungal populations
75 found on major widespread crops, which due to their vast population sizes, benefit
76 from a high input of novel variation by mutations. These new pathogen genotypes
77 can rapidly spread through genetically uniform host populations. In other words,
78 the high evolvability of some fungi during biological invasions is not necessarily
79 realized through their high standing genetic variation at the point of entry (cf.
80 Fisher's Fundamental Theorem (Price, 1972)). Rather, many fungi are highly
81 potent biological invaders due to input of *de novo* allelic and genotypic variation
82 every generation, which is a consequence of their high potential for gene flow,

83 recombination and mutation. These drivers of genetic variation play a crucial role
84 in the co-evolutionary arms race between fungal pathogens and their hosts. A
85 sudden increase in these drivers can shift the co-evolutionary balance, and these
86 effects can be particularly severe in an eco-agriculture setting with genetically
87 relatively uniform host plants and animals (Van Oosterhout, 2021).

88 Multiple populations of the same fungal species can coexist on one host whilst
89 competing for limited resources (Bueno-Sancho et al., 2017; Fournier, Gladieux,
90 & Giraud, 2013; Hartmann, McDonald, & Croll, 2018; Hubbard et al., 2015;
91 Persoons et al., 2017; Silva, Várzea, Paulo, & Batista, 2018; Stauber et al., 2020;
92 Thierry et al., 2020; Vieira, Silva, Várzea, Paulo, & Batista, 2018). In the absence
93 of temporal, spatial or habitat barriers, coexistence on the same host may foster
94 genetic exchanges between fungal lineages. If coexisting populations represent
95 previously geographically isolated lineages that have not evolved strong pre- or
96 postzygotic barriers, such introgression can rapidly generate novel genotypic
97 variation. In turn, this can increase the amount of phenotype variation – a
98 phenomenon known as transgressive segregation – which provides more novel
99 substrate for natural selection (Nichols et al., 2015).

100 Admixture between multiple coexisting populations can also lead to a so-
101 called bridgehead effect, in which highly adapted lineages emerge through
102 recombination among propagules established in an area of first introduction
103 (Bertelsmeier & Keller, 2018; Dutech et al., 2012; Stauber et al., 2020). Ongoing
104 fungal invasions offer a unique opportunity to learn about the ecology and
105 evolution of biotic interactions in human-altered ecosystems (Gladieux et al., 2014;
106 Parker & Gilbert, 2018; Thrall, Hochberg, Burdon, & Bever, 2007; Thrall et al.,

107 2011), and such studies are important to assess the risks posed by pathogens to crop
108 in agriculture.

109 Anthracnose, mainly associated with the fungus *Colletotrichum truncatum*
110 (Hyde et al., 2009), is one of the most prominent foliar diseases of soybean. This
111 ascomycete is seed-transmitted and can infect more than 460 plant species,
112 including important crops in the Fabaceae and Solanaceae families (Cannon,
113 Damm, Johnston, & Weir, 2012; Damm, Woudenberg, Cannon, & Crous, 2009;
114 Weidemann, TeBeest, & Cartwright, 1988). In Brazil, the worldwide leader in
115 soybean production, previous population genetic studies showed that *C. truncatum*
116 is a recently introduced invasive species structured into three highly divergent
117 clusters coexisting in soybean fields. This suggests there have been multiple
118 introductions from distinct source populations, which are yet to be identified
119 (Rogério, Gladieux, Massola, & Ciampi-Guillardi, 2019).

120 *Colletotrichum* genus are predominantly observed in the asexual phase, and
121 the sexual morph is rarely identified for most species (Barcelos, Pinto,
122 Vaillancourt, & Souza, 2014; Talhinhos & Baroncelli, 2021). Mating system of this
123 group does not follow the usual bipolar self-incompatibility MAT1-1/2 system
124 (Crouch et al., 2014), typically found in ascomycetes. All species that have been
125 studied to date have only the single mating type MAT1-2, however, the loss of one
126 idiomorph does not affected the sexual capabilities of the species, which has led to
127 the development of hypotheses to describe this singular mating behavior (Menat,
128 Banniza, Cabral, Vijayan, & Wei, 2012; Vaillancourt et al., 2000; Wilson, Lelwala,
129 Taylor, Wingfield, & Wingfield, 2021).

130 *C. truncatum* is a homothallic species what mean that are capable of sexual
131 reproduce without a mating partner. Although sexual reproduction has not been

132 described until date to this species, alternative sexual mechanisms are described to
133 other species (Roca, Davide, Mendes-Costa, & Wheals, 2003; Rosada et al., 2010;
134 Souza-Paccola, Fávaro, Casela, & Paccola-Meirelles, 2003; Vaillancourt et al.,
135 2000), and such processes also may be play role in *C. truncatum* reproduction.
136 Some studies reported population genetic signatures of sexual recombination in *C.*
137 *truncatum* (Diao et al., 2015; Katoch, Sharma, Padder, & Sharma, 2017; (Diao et
138 al., 2015; Katoch, Sharma, Padder, & Sharma, 2017; Rogério, Gladieux, Massola,
139 & Ciampi-Guillardi, 2019) signaling that recombination events may be occurring
140 and playing an important role in the genetic diversity of this pathogen.

141 Here, we use whole-genome **sequence** and a population genomics approach
142 to characterize the genetic makeup and infer the evolutionary history of *C.*
143 *truncatum* causing soybean anthracnose in Brazil. We document evidence of
144 extensive introgression between three **sympatric** lineages that have invaded
145 Brazilian soybean. Our study highlights the risk that Brazilian *C. truncatum* may
146 represent as a bridgehead for future invasions of soybean-producing areas,
147 facilitating admixture between the three lineages (as well as with any unsampled
148 lineages and possible future immigrant lineages). We discuss why fungi have the
149 means to become potent biological invaders of crops, arguing this is due to their
150 ability to rapidly generate novel genetic variation, in combination with their high
151 propagule pressure accomplished through two **modes** of asexual reproduction. We
152 furthermore discuss why the large biomass of relative uniform crops provides the
153 motive, and the bridgehead populations that enable genetic introgression the
154 opportunity for such biological invasions.

155

156 **MATERIALS AND METHODS**

157 **Fungal isolates, DNA extraction and genome sequencing**

158 We used 18 isolates of *Colletotrichum truncatum* from naturally infected
159 commercial soybean fields from Mato Grosso (MT) and Goiás (GO) states in Brazil.
160 Isolates **coexisting in the fields sampled** were randomly selected from the three genetic
161 clusters (C1, C2, and C3) previously identified based on the population genetics analysis
162 of microsatellite variation (Rogério, Gladieux, Massola, & Ciampi-Guillardi, 2019)
163 (Table 1). **Information about isolation and culture collection can be found in the study**
164 **mentioned.** Fungal genomic DNA was extracted using Wizard Genomic DNA
165 Purification kit (Promega) from fresh mycelium grown on potato dextrose liquid medium
166 (Difco). Paired-end libraries were prepared and sequenced on Illumina HiSeq2000
167 (2x150 bp, insert size ~ 500 pb) by Genewiz (South Plainfield, USA). The raw reads
168 were deposited at the NCBI/Genbank under the Sequence Read Archive (SRA) accession
169 numbers SAMN13196067 to SAMN13196084 (Table S1).

170

171 **Read mapping and SNP calling**

172 Read quality was checked using FASTQC
173 (<https://www.bioinformatics.babraham.ac.uk/projects/fastqc/>). Raw Illumina reads were
174 trimmed for adapter contamination and bases **with a Phred score** smaller than 30 were
175 removed using CUTADAPT v1.16 software (M. Martin, 2011). Reads were mapped
176 against the reference genome of the previously assembled isolate CMES1059 (Rogério
177 et al., 2020) using BWA-MEM v0.7.15 (options -n=5) (Li & Durbin, 2009). Alignments
178 were sorted with SAMTOOLS v1.3 (Li et al., 2009), and reads with mapping quality below
179 30 were removed. Duplicates **reads** were removed using PICARD v2.7
180 (<http://broadinstitute.github.io/picard/>). Single nucleotide polymorphisms (SNPs) and
181 indels were called using the HAPLOTYPECALLER module from the Genome Analysis

182 Toolkit v4.0.12 (GATK) (McKenna et al., 2010), with the option *-emitRefConfidence*
183 GVCF. The gVCF files listing variants were merged using COMBINEGVCFs and
184 genotyped using GENOTYPEGVCFs. Monomorphic sites were included using the
185 argument *include_nonvariantsites*. High confidence SNPs were identified using GATK's
186 VARIANTFILTRATION module, following GATK's best practices
187 (http://www.broadinstitute.org/gatk/guide/best_practices), with parameters: QD < 2.0
188 (Variant Quality), FS > 60.0 (Phred score Fisher's test), MQ < 40.0 (Mapping Quality),
189 MQRankSum < 12.5 (Mapping Quality of Reference reads vs alternative reads) and
190 ReadPosRankSum < 8.0 (Distance of alternative read from the end of the reads).

191

192 **Population structure**

193 We used SPLITSTREE v4 (Huson & Bryant, 2006) to visualize relationships
194 between isolates in a phylogenetic network based on pseudoassembled genomic
195 sequences (SNPs in vcf format were converted into a fasta file using the reference
196 sequence as a template) generated from the tables of SNPs with the reference sequence
197 as a template. We also used the pairwise homoplasy index (PHI) test implemented in
198 SPLITSTREE to test the null hypothesis of clonality; recombination is expected to result in
199 exchangeable sites within lineages.

200 Population structure was analyzed using methods optimized for the analysis of
201 large datasets and which do not assume Hardy-Weinberg equilibrium. We performed
202 principal components (PCA) and discriminant analysis of principal components (DAPC)
203 in R with the package ADEGENET v2.0 (Jombart & Ahmed, 2011), using the
204 FIND.CLUSTERS function. DAPC is a non-model-based method using PCA as a prior step,
205 which provides a description of clusters using discriminant functions. We retained the
206 first 20 principal components. DAPC identifies an optimal number of genetic clusters

207 that best describes the data by running a k-means clustering algorithm and comparing the
208 different clustering solutions using the Bayesian Information Criterion (BIC). Population
209 structure was also analyzed using the software sNMF (Frichot, Mathieu, Trouillon,
210 Bouchard, & François, 2014) which estimates ancestry coefficients based on sparse non-
211 negative matrix factorization and least squares optimization. We calculated ancestry
212 coefficients for 2 to 10 ancestral populations (K) using 100 replicates for each K . The
213 preferred number of K was chosen using a cross-entropy criterion based on the prediction
214 of masked haplotypes to evaluate the error of ancestry estimation. All clustering analyses
215 were based on biallelic SNPs without missing data.

216

217 **Diversity and divergence**

218 Polymorphism and divergence statistics were computed in 10 kb non-overlapping
219 windows using the SCIKIT-ALLEL v1.0.2 Python package (Miles & Harding, 2017).
220 Summary statistics were plotted for the 10 largest contig using CIRCOS v0.67 software
221 (Connors et al., 2009).

222

223 **Recombination analyses**

224 Recombination events were analyzed using the software RDP4 (Martin, Murrell,
225 Golden, Khoosal, & Muhire, 2015) implementing seven independent detection
226 algorithms: RDP (Martin & Rybicki, 2000), GENECONV (Padidam, Sawyer, & Fauquet,
227 1999), BOOTSCAN (Martin, Posada, Crandall, & Williamson, 2005), MAXCHI (Smith,
228 1992) CHIMAERA (Posada & Crandall, 2002), SiSCAN (Gibbs, Armstrong, & Gibbs,
229 2000) and 3SEQ (Boni, Posada, & Feldman, 2007). Whole sequences of the ten largest
230 contigs were scanned using the default settings for the window size. Tests were
231 conducted using a critical value $\alpha = 0.05$ and p-values were Bonferroni corrected for

232 multiple comparisons of sequences. The evidence for a recombination signal was
233 considered to be strong if it was found to be significant with three or more detection
234 methods. Only events for which the software identified the parental sequences (i.e., no
235 ‘unknowns’) without ambiguous start and end position of the recombination block were
236 considered.

237 We used POPLDDECAY version v3.4 (Zhang, Dong, Xu, He, & Yang, 2019) to
238 investigate the patterns of linkage disequilibrium decay within *C. truncatum* genetic
239 groups as coefficient of linkage disequilibrium (r^2) (Hill & Robertson, 1968) calculated
240 for all pairs of SNPs less than 300 kb apart. For this, we used biallelic SNPs, excluding
241 missing data and sites with minor allele frequencies below 10%.

242

243 **Genome scan for signature of genetic exchanges**

244 We used CHROMOPAINTER v0.0.4 (Lawson, Hellenthal, Myers, & Falush, 2012)
245 for probabilistic chromosome painting to infer recent shared ancestry between *C.*
246 *truncatum* lineages. This method “paints” individuals in “recipient” populations as a
247 combination of segments from “donor” populations, using linkage disequilibrium
248 information for probability computation and assuming that linked alleles are more likely
249 to be exchanged together during recombination events. We ran three separate analyses,
250 each considering one particular *C. truncatum* lineage as a collection of haplotypes to be
251 painted, and all lineages as donors. The recombination scaling constant N_e and emission
252 probabilities (μ) were calculated as averages weighted by contigs’ length determined by
253 LDHAT (Auton & McVean, 2007). Estimates of these parameters for each lineage were
254 obtained by running the expectation maximization algorithm with 200 iterations. These
255 analyses were based on biallelic SNPs dataset without missing data.

256 Fine-scale admixture between *C. truncatum* lineages also was analyzed using
257 the software HYBRIDCHECK (Ward & van Oosterhout, 2016), which uses a sliding
258 window to scan for sudden changes in nucleotide divergence between sequences, thus
259 identifying potential genetic exchanges where nucleotide divergence is significantly
260 lower. The similarities were visualized through a plot employing the primary colors red,
261 green, and blue, using the 100 bp windows based on the proportion of SNPs shared
262 between the pairwise sequences, with a stepwise increment of 1 bp. In cases where all
263 SNPs are shared between just two of the three lineages, the hybrid color is an exact 50%
264 mix of two primary colors. Hence, yellow, purple, and turquoise colors pinpoint regions
265 of possible recent genetic exchange between two sequences. We carried out this analysis
266 on a triplet involving one isolate representative of each lineage: isolates LFN0297
267 (lineage C1), LFN0318 (lineage C2), and LFN0308 (lineage C3) for the ten largest
268 contigs.

269 HYBRIDCHECK and CHROMOPAINTER identify genomic regions of shared
270 ancestry, and such signal can be caused either by genetic introgression, or by incomplete
271 lineage sorting (Durand, Patterson, Reich, & Slatkin, 2011). In order to differentiate
272 between genetic introgression and incomplete lineage sorting, we also used
273 HYBRIDCHECK to estimate the age of recombinant regions. If the genomic region
274 coalescence before the split of the species (or lineages), the signal is consistent with
275 incomplete lineage sorting. However, if the coalescence event is dated after the speciation
276 event (or after the bifurcation of the lineages in the tree), the genetic exchange has
277 occurred after the divergence. In the latter case, the signal is consistent with genetic
278 introgression after hybridization (Jouet, McMullan, & Van Oosterhout, 2015).
279 Recombination blocks were then dated assuming a strict molecular clock with a mutation
280 rate of 10^{-8} per generation, assuming a generation time of one year.

281

282 **Functional enrichment**

283 To characterize genes present in the introgressed regions between *C. truncatum*
284 lineages, we extracted the transcripts and proteins from the corresponding regions in the
285 reference genome (Rogério et al., 2020) using GFFREAD (Pertea & Pertea, 2020) **using**
286 **the ten largest contigs as reference background. For the enrichment analysis the** GO terms
287 were assigned from re-annotated transcripts using BLAST2GO (Conesa et al., 2005)
288 against the NCBI non-redundant database **and used as input to the** TOPGO **R package**
289 **(Alexa & Rahnenfuhrer 2021)**. We used SIGNALP v5.0 (Armenteros et al., 2019) to
290 identify secreted proteins.

291

292 **Demographic inferences**

293 To infer the evolutionary history of the genetic lineages we used the Python
294 package DADI (Gutenkunst, Hernandez, Williamson, & Bustamante, 2009). The method
295 implemented in DADI infers demographic parameters based on a diffusion approximation
296 to the site frequency spectrum (SFS). The Python script EASYSFS.PY (available at
297 <https://github.com/isaacovercast/easySFS>) was used to convert the VCF file into a three-
298 dimensional joint site frequency spectrum (3D-JSFS). The SFS was folded because no
299 appropriate outgroup was available. We compared twelve demographic models including
300 strict isolation, isolation with migration (asymmetrical migration rates), and isolation
301 with population size changes, with four possible topologies, using the demographic
302 modeling workflow (*dadi_pipeline*) from Portik et al. 2017 (Fig. S1). For each model,
303 we performed four rounds of optimizations; for each round, we ran multiple replicates
304 and used parameter estimates from the best scoring replicate (highest log-likelihood) to
305 seed searches in the following round. We used the default settings in *dadi_pipeline* for

306 each round (replicates = 10, 20, 30, 40; maxiter = 3, 5, 10, 15; fold = 3, 2, 2, 1), and
307 optimized parameters using the Nelder-Mead method (*optimize_log_fmin*). We used the
308 optimized parameter sets of each replicate to simulate the 3D-JSFS, and the multinomial
309 approach was used to estimate the log-likelihood of the given the model. We assessed
310 the model's goodness-of-fit by maximizing the model likelihood and visual inspection of
311 the residuals between the site frequency spectra generated by the inferred model and the
312 real data. Conventional bootstrapping (100 replicates) was carried out to estimate
313 parameter uncertainties, by dividing data into 100 independent chunks of 1 Mb in which
314 SNPs were sampled with replacement.

315

316 **RESULTS**

317 **Population structure and levels of genetic variation**

318 Read mapping and variant calling for 18 isolates of *C. truncatum* submitted to
319 whole-genome sequencing identified 2,220,191 biallelic Single Nucleotide
320 Polymorphisms (SNPs), distributed across 128 contigs (see Table S1 to
321 sequencing statistics). To assess population subdivision and to visualize relationships
322 among isolates we built a neighbor-net network with SPLITSTREE based on the full set of
323 SNPs. This phylogenetic network revealed three groups, henceforth referred to as
324 "lineages" C1, C2 and C3 (Fig. 1A). Lineage C1 was connected to the rest of the dataset
325 by a long, non-reticulated branch consistent with relatively long-term genetic isolation.
326 Lineage 2 and 3 were connected by branches showing extensive reticulations (looping in
327 the network) indicating a history of recombination or incomplete lineage sorting (Fig.
328 1A). In a Discriminant Analysis of Principal Component (DAPC) modelling $K=2$ to
329 $K=10$ populations, the Bayesian Information Criterion monotonously decreased with
330 increasing K , preventing clear choice of a best supported model, but the composition of

331 clusters identified at $K=3$ matched what was observed in the neighbor-net network (Fig.
332 1B and Fig. S2). Likewise clustering by sparse non-negative matrix factorization
333 algorithms, as implemented in the sNMF method, identified $K = 3$ as the best supported
334 model based on cross-entropy. The ancestry coefficients estimated with sNMF revealed
335 essentially the same pattern of population subdivision as the DAPC and neighbor-net
336 network. However, three isolates (LFN0318, LFN0217, and LFN0150) shared ancestry
337 in two clusters, suggesting admixture between lineages (Fig.1C).

338 Nucleotide diversity was nearly twice as high in C3 than in C2, and it was more
339 than one order of magnitude higher in C2 than in C1 (C3: $\pi=0.0113/\text{bp}$; C2: $\pi=0.0062/\text{bp}$;
340 C1: $\pi=0.0002/\text{bp}$; Table 2; Fig.2). In lineage C2, regions of relatively high nucleotide
341 diversity were interspersed with tracts of low diversity (Fig. S3). Tajima's D values were
342 either close to zero, or they were negative in the three lineages (C1: $D=0.008$; C2: $D=-$
343 0.380 ; C3: $D=-0.177$; Table 2); a negative value is consistent with population expansion
344 after a recent bottleneck or founder event. Absolute divergence (d_{xy}) among lineages was
345 similar between the three pairs of lineages ($d_{xy}=0.018/\text{bp}$ between C1 and C2, $0.018/\text{bp}$
346 between C2 and C3, and $0.015/\text{bp}$ between C1 and C3; Fig. 2).

347

348 **Footprints of recombination**

349 Recombination was analyzed in the ten largest contigs of each lineage, covering
350 ~30% of the reference assembly. We detected a total of 375 recombined blocks using
351 RDP4 software on the contigs analyzed, with stretches of nucleotide similarity across
352 lineages distributed in a block-like structure (Table S2). Recombination rates
353 differed significantly across lineages; based on the total of 375 recombination events,
354 lineage C3 was found to have received the highest number of recombination events
355 ($n=203$), followed by C2 ($n=148$), and with C1 receiving significantly fewer events

356 (n=24) (Randomization test: $p < 10^{-6}$; Table S2). In this analysis, we counted the number
357 of cases in which C1, C2 or C3 was the recombinant in Table S2. Analyses of linkage
358 disequilibrium showed that LD decayed to half of its maximum value in less than 1kb in
359 lineages C2 and C3, while LD decay was markedly slower and more jagged in C1 (Fig.
360 3). The large number of recombination events would have homogenized the nucleotide
361 diversity and broken up any LD blocks in C2 and C3, resulting in a smooth LD decay. In
362 contrast, the LD decay is more erratic in C1 because the few recombination events have
363 not managed to break-up all LD blocks. **We also tested for an alternative explanation; the**
364 **smaller number of SNPs in C1 could have introduced stochasticity in the LD values as**
365 **these values are calculated on fewer data. However, subsampling a smaller number of**
366 **random SNPs from C2 and C3 to give them the same nucleotide diversity as C1 did not**
367 **produce the erratic pattern of LD-decay. Hence, we conclude that the level of sexual**
368 **recombination within C1 is lower than that of both other lineages.** Finally, the PHI test
369 rejected the null hypothesis of clonality in all three lineages ($P < 0.001$).

370

371 **Genome scan for signatures of genetic exchanges**

372 Probabilistic chromosome painting revealed genomic regions of shared ancestry
373 between lineages, with shared fragments of size longer than the longest contig in the
374 reference genome (2.37 Mb) (Fig. S4). Regions of shared ancestry were not strictly
375 restricted to the three isolates (LFN0318, LFN0217, and LFN0150) previously detected
376 by sNMF (Fig.1). For lineages C1 and C2, the majority of mutations were assigned to
377 self (i.e., to their cluster of origin), but in some contigs relatively large regions were
378 assigned to other lineages. In lineage C3, mutations tended to have non-zero membership
379 probabilities in multiple clusters. This implies that these polymorphisms are shared
380 across multiple lineages, probably reflecting the extremely high recombination rate

381 shown by this lineage. However, also in lineage C3, some regions were clearly assigned
382 to lineage C1 and self. For contig66, the isolate LFN0318 – a representative of lineage
383 C2 – shared high genetic similarity with lineage C3, consistent with many recent genetic
384 exchanges between these lineages.

385 Further analyses using HYBRIDCHECK revealed a mosaic-like genome structure
386 with well-defined blocks of high nucleotide similarity (Fig. S5). For contig66, relatively
387 few short blocks of high similarity were detected between lineage C1 (using LFN0297
388 as the representative isolate) and the two other lineages (spanning from 1.4 to 1.6 Mb),
389 while large blocks were detected between lineage C2 and C3 (using LFN0318 and
390 LFN0308 as the representative isolates, respectively) (Fig. 4). Assuming that the contigs
391 analyzed are representative of the rest of the genome, the proportion of genome
392 introgression between lineages varied markedly: 2.4% between C1 (LFN0297) and C2
393 (LFN0318); 12.7% between C1 (LFN0297) and C3 (LFN0308); and 28.7% between C2
394 (LFN0318), and C3 (LFN0308) (Table 3 and Table S3).

395 In order to discriminate between incomplete lineage sorting and hybridization
396 after secondary contact, we dated regions of high nucleotide identity between *C.*
397 *truncatum* lineages detected by HYBRIDCHECK. The age estimates of recombinant blocks
398 along the 10 largest contigs revealed recent introgression events. The most recent
399 hybridization event was dated back to 6,100 years before present, assuming a generation
400 time of one year (Table S4). Variation in the age of introgression events was however
401 extensive, most likely because of a lack of closest-related donors for all recombinant
402 regions detected (see Jouet, McMullan, & Van Oosterhout, 2015). The most recent events
403 were more likely to reflect ongoing genetic exchanges. The older events are more
404 consistent with incomplete lineage sorting, or alternatively, the coalescence time may not

405 accurately reflect the timing of genetic introgression because the “true” donors have not
406 been sampled.

407 In the introgression regions, we identified 357 genes between lineage C1
408 (LFN0297) and lineage C2 (LFN0318), 389 genes between lineage C1 (LFN0297) and
409 lineage C3 (LFN0308), and 584 genes between lineage C2 (LFN0318) and C3
410 (LFN0308). The introgressed regions include many secreted protein-encoding genes
411 (between 38 and 62) (Table S5), including proteases and hydrolases which are known
412 virulence-associated factors in pathogens (Monod et al., 2002; Soanes, Richards, &
413 J.Talbot, 2007). However, these regions are not significantly enriched for those genes
414 (binomial test $p > 0.05$), and hence, we must conclude that genetic exchanges between
415 these lineages are not more likely to involve genomic regions with virulence genes. Gene
416 Ontology (GO) analysis revealed enrichment of GO terms between lineages, and these
417 results are reported in Table S6.

418

419 **Demographic inferences**

420 To infer the demographic history of the three genetic lineages of *C. truncatum*,
421 we compared three scenarios of isolation with or without migration for the four possible
422 branching orders among lineages, using a diffusion approximation to the SFS
423 implemented using DADI (We compared $3 \times 4 = 12$ models compared in total). Note that
424 in the context of the DADI analysis, the term “migration” is similar to “genetic
425 introgression” in the recombination analysis. Likelihood ratio tests indicated that the
426 model with trifurcating lineages (topology 1) and asymmetrical migration was the most
427 supported (Fig. S6). These results corroborate our recombination analyses. To convert
428 demographic parameter estimates to physical units, we estimated the ancestral population
429 size (N_{AB1}) (Fig.5). We used the population mutation rate $\theta = 4N_{AB1}\mu L$, where μ was

430 assumed to be approximately $1e-8$ per generation (Lynch, 2010) and L was the genome
431 size (~ 55.1 Mb). This ancestral population size was then used to transform time estimates
432 from DADI (in units of $2N_{AB1}$) into calendar years. Divergence was estimated to have
433 initiated about 960,000 years ago, considering a generation time of a year. The lowest
434 population size was estimated for lineage C1, consistent with a severe bottleneck
435 ($n_{u1}=2085$ individuals), and/or the least input of genetic variation through recombination
436 or genetic introgression.

437 The migration rate into C3 ($m_{13}=1.05e-6$, $m_{23}=2.25e-6$) was higher than in all
438 other directions, displaying one to two times higher order of magnitude. Furthermore, the
439 rate of migration into C1 (m_{21} and m_{31}) was lower than in all other directions. This
440 suggests lineage C3 as the higher receptor of migration. These demographic analyses
441 thus support the recombination and introgression analyses reported above.

442

443 **Discussion**

444 The fungus *Colletotrichum truncatum* is an invasive pathogen on soybean crops
445 in Brazil that causes severe yield losses. We used a population genomics approach to
446 characterize the genetic makeup and infer the evolutionary history of *C. truncatum* using
447 isolates representing two important regions of soybean production in Brazil. We showed
448 that Brazilian *C. truncatum* is subdivided into three genetic lineages. These lineages
449 possess different levels of genetic variation, which could reflect differences in the
450 magnitude of bottlenecks associated with introduction events. A non-exclusive
451 alternative hypothesis is that such differences have been caused by variation in the levels
452 of recombination (i.e., genetic exchanges within a lineage) and/or genetic introgression
453 (genetic exchanges between lineages). All our recombination analyses supported that the
454 lineage C3 with the highest nucleotide diversity was most affected by such genetic

455 exchanges, and that a larger number of these events also affected more genes.
456 Conversely, C1 with the lowest nucleotide variation and the highest level of linkage
457 disequilibrium was least affected by genetic exchanges. Furthermore, these conclusions
458 are corroborated by our demographic analysis which showed that migration (or gene
459 flow) into C1 (m21 and m32) is lower than into all other directions, and that migration
460 into C3 was the highest. Next, we will discuss the evolutionary genomics of *C.*
461 *truncatum*, the significance of the Brazilian bridgehead population, and the potential of
462 fungal pathogens to evolve into invasive species.

463

464 *Evolutionary genomics of C. truncatum*

465 Our clustering analyses supported the existence of three lineages, in agreement
466 with the pattern of population subdivision previously detected based on multilocus
467 microsatellite typing (Rogério, Gladieux, Massola, & Ciampi-Guillardi, 2019). These
468 phylogenetically equidistant lineages are characterized by markedly different levels of
469 genomic standing variation. **Genome-wide analyses of variability showed that lineage C1**
470 **is genetically depauperate and reproductively more isolated.** Lineage C1 was almost free
471 from introgression, whilst between ~10 to 30% of assembled contigs of lineage C2 and
472 C3 comprised recombinant (or introgressed) regions. The low level of genetic variation
473 in C1 is consistent with the low number of recombinant blocks (only 6.4%), but it can
474 also indicate a more recent introduction into the country. It means that C1 lineage may
475 simply not have had the opportunity to engage in many genetic exchanges yet. Such
476 recent invasions could be associated with contaminated or infected soybean seeds
477 imported from the U.S. during the 1960s and 1970s (Arantes and Miranda 1993;
478 Hirimoto and Vello 1986; Wysmierski and Vello 2013). Regions of low nucleotide
479 diversity in this lineage corresponded with negative Tajima's D values, which is

480 consistent with rapid population expansion after a recent founder event. These
481 observations lend further support to our demographic inference of a recent invader, the
482 limited amount of introgression, and the lower effective population size (N_e) of C1
483 lineage.

484 By contrast, lineage C3 showed a significantly higher level of genetic diversity,
485 which may have been generated over time by genetic recombination and its much larger
486 N_e (note that the large N_e estimate may simply be a consequence of the ample genetic
487 variation that has introgressed into this lineage). This lineage may have already been
488 present in Brazil, prior to the introduction of soybean. It is possible that this lineage may
489 have been infecting other host species, such as lima bean and weeds, as proposed by
490 earlier studies (Rogério, Gladieux, Massola, & Ciampi-Guillardi, 2019; Tiffany &
491 Gilman, 1954). If C3 was the first lineage that established itself, it would also have had
492 more opportunity for recombination and genetic introgression than both other lineages,
493 which could have augmented its genetic diversity. Such monopoly effect may also have
494 given this lineage a head-start, both in the co-evolutionary arms race with its host (Van
495 Oosterhout, 2021), as well in the competition with other lineages.

496 We found evidence of a history of recombination both within and between
497 lineages, applying a combination of integrated approaches. Fine-scale admixture
498 mapping revealed that introgression occurred between the *C. truncatum* lineages
499 coexisting in sympatry despite their relatively deep genomic divergence. The inference
500 of individual ancestry coefficients using probabilistic chromosome painting detected
501 large genomic regions of shared ancestry among the genetic lineages, suggesting
502 relatively recent hybridization. History of recombination and genetic introgression was
503 also supported by the analyzes with HYBRIDCHECK, which detected introgressed blocks
504 that differed markedly in age. By estimating the coalescence time of introgressed regions,

505 we found some events have occurred as recently as 6,100 years ago. This suggests that
506 hybridization between *C. truncatum* lineages is a relatively recent - if not ongoing -
507 process. In this analysis, we used a sexual generation time of a year, which is typical in
508 plant pathogenic fungi from temperate areas. However, considering the climate
509 conditions of Brazil, the generation time may be much shorter than one year, and hence,
510 we may have overestimated the age of introgression events. Furthermore, it is unlikely
511 we sampled the actual parental sequence, which would cause a further overestimation of
512 the age. (when identifying the wrong parental sequence, the SNPs that differentiate the
513 parent and the recipient sequence are assumed to have accumulated since recombination
514 took place, erroneously placing the recombination event further in the past). In other
515 words, hybridization events and genetic exchanges may be considerably more recent than
516 our estimate. This is a potentially systematic bias typical for recombination studies, and
517 this can be corrected for by broader (or more intense) sampling.

518 Although our analyses identified recent introgression between lineages, a
519 **substantiate** proportion of the shared ancestry observed between lineages appear to be
520 caused by incomplete lineage sorting (or alternatively, we overestimated the age). In such
521 cases, the recombination blocks pre-date the lineage divergence (Durand et al., 2011;
522 McMullan et al., 2015), which implies that they are shared ancestral polymorphisms, or
523 that the genetic exchanges occurred before the split of the lineages. **Consistent with this**
524 **latter interpretation, the oldest blocks were also the shortest, which implies that they have**
525 **been eroded by subsequent recombination and mutations, and hence, that these represent**
526 **genetic exchanges predating the split of the lineages.** *Colletotrichum truncatum* genomes
527 therefore appear to be mosaics of distinct gene genealogies with markedly varied
528 coalescence times. Alternatively, we may not have captured all extant lineages, which
529 would have overestimated the introgression events. Future studies with a more

530 comprehensive sampling may be able to shed further light on this. Next, we discuss our
531 finding in the context of fungi as invasive species.

532

533 *Evolutionary genomics of fungi as invasive species*

534 Fungal reproductive biology is **conducive** for genetic exchange, and such
535 recombination events could both **result** from sexual reproduction or parasexual events
536 via hyphal anastomosis. The latter mode has already been described for other
537 *Colletotrichum* species (Roca et al., 2003; Rosada et al., 2010; Souza-Paccola et al.,
538 2003; Vaillancourt et al., 2000). Given that such genetic exchanges were found in all
539 genomes – i.e., no pure genomes were found – introgression is likely to have augmented
540 both the genetic diversity and the fitness of these hybrid genotypes. Therefore, we would
541 conclude that adaptive introgression may have enhanced the evolutionary potential of *C.*
542 *truncatum* during its invasion. However, when we tested this hypothesis, we did not find
543 significant enrichment of secreted protein-encoding genes in the introgressed regions. In
544 hindsight, this may not be surprising; in a coevolutionary arms race, genetic novelty at
545 single virulence gene introduced by recombination could provide a selective advantage
546 that helps the recombinant lineage to establish itself (Van Oosterhout, 2021). Indeed,
547 specific targets are likely to be under positive selection, rather than the total number of
548 introgressed genes (Aguileta, Refrégier, Yockteng, Fournier, & Giraud, 2009). In other
549 words, our study may not have discovered “the smoking gun”, but we have established
550 “the means”, i.e., the large number of secreted protein-encoding genes that are exchanged
551 during genetic introgression, which are possible co-evolutionary targets for selection.
552 This implies that introgression can provide the genetic variation required in a host-
553 parasite arms race, **although we do not have the phenotypic data to support this**
554 **hypothesis.**

555 The level of phylogenetic divergence among lineages, coupled with the
556 demographic modeling carried out in this study, enables us to infer that the *C. truncatum*
557 lineages significantly diversified before their joint introduction into Brazil. Based on the
558 ancient signature of some of the recombination events, it is possible that genetic
559 exchanges have occurred during the divergence process. These genetic exchanges would
560 have prevented the accumulation of intrinsic postzygotic barriers, which underpin
561 reproductive isolation in many recently diverged species (Bomblies et al., 2007; Lee et
562 al., 2008; Masly & Presgraves, 2007). Another possibility is that they initially evolved in
563 allopatry, but their later introduction in the same areas may have provided opportunities
564 for secondary contact and hybridization, preventing the accumulation of postzygotic
565 barriers. In summary, the absence of reproductive isolation may have provided ample
566 evolutionary opportunities after secondary contact, allowing for hybridization between
567 diverged lineages.

568 Introduced populations can overcome consequences of low genetic variation from
569 founders, for instance, through the purging of deleterious alleles during bottlenecks
570 events, and via the fixation of *de novo* beneficial mutations from standing variation
571 (Estoup et al., 2018; Frankham, 2005; Schrieber & Lachmuth, 2017). The genetic and
572 environmental homogeneity found in soybean fields is thought to favor a huge census
573 size of invasive fungal populations. Furthermore, without genetic diversity in the host,
574 pathogens can rapidly spread and fix *de novo* evolved adaptations, overcoming new
575 resistant varieties of crops and fungicides applications. Hybridization could be
576 particularly important in this context because admixture could promote adaptation by
577 rapidly creating novel allelic combinations (Hessenauer et al., 2020; McMullan et al.,
578 2015; Nader et al., 2019). We propose that admixture has elevated the amount of
579 genotypic variation generated by genetic introgression. In turn, this could have increased

580 the amount of phenotypic variation due to transgressive segregation, providing novel
581 substrate for natural selection (Nichols et al., 2015).

582 Bridgehead population may enhance the adapted invasive potential of species by
583 enabling genetic exchanges between diverged lineages in the areas of first introduction.
584 In this scenario, bridgehead populations may acquire new traits increasing the probability
585 of successful establishment and further spread relative to native population (Bertelsmeier
586 & Keller, 2018). We hypothesize that the admixture between *C. truncatum* lineages may
587 lead to a bridgehead population, producing a highly adapted invasive population.
588 Although the bridgehead effect has been proposed as a potential explanation for many
589 successful biological invasions (Gau, Merz, Falloon, & Brunner, 2013; Leduc et al.,
590 2015; van Boheemen et al., 2017) there is currently no clear empirical support for this
591 hypothesis (but see Simon et al., 2011 for an example of an invasive fish species). To
592 contain biological invasions, vigilance for invasive bridgehead populations is needed
593 since they have the potential to generate new introductions (Bertelsmeier & Keller, 2018)
594 and increase the adaptive evolutionary potential through genetic reassortment during
595 hybridization.

596 Our study reinforces the practical applications of population genomics in preventing
597 or curtailing, pathogen dissemination by supporting early interventions to limit economic
598 damage (Stam et al., 2021). The Brazilian *C. truncatum* may represent a risk as a
599 bridgehead for future invasions of soybean-producing areas. Our data highlights the
600 inherent vulnerability of genetically uniform crops in the agro-ecological environment,
601 particularly when faced with pathogens that can take full advantage of the opportunities
602 offered by an increasingly globalized world. Some fungi have “The Means, Motive and
603 Opportunity” to become invasive pathogens of crops. Many fungal pathogens possess the
604 means in the form of a high propagule pressure through two modes of asexual

605 reproduction, as well as the ability to rapidly generate novel genotypic variation,
606 particularly through genetic introgression. Some fungi also have a motive, given the large
607 biomass of genetically near-uniform crops that are their natural host plants. Few species
608 also have the opportunity in the form of bridgehead populations that enable the genetic
609 exchange that fuel the co-evolutionary arms race. Invasive crop pathogens, like *C.*
610 *truncatum*, have “The Means, Motive and Opportunity” to pose the greatest risk to future
611 food security. Population genomics can help identify pathogens that pose such risk,
612 thereby helping to inform control strategies to better protect crops in the future.

613

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621

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933 **Data accessibility**

934 DNA sequences: Short Read Archive Accession in Table S1 (Supporting information)
935

936 **Author Contributions**

937 M.C.G., and N.S.M.J conceived and designed the research; F.R., and M.C.G collected
938 the samples; S.C.A. obtained genomic data; F.H.C., G.K.H., and G.R.A.M. performed
939 genetic analysis; F.R., C.V.O., and P.G. analyzed the data and wrote the manuscript.

940

941 **Figure captions**

942 **Figure 1.** Population subdivision of *Colletotrichum truncatum*. (A) Neighbor-Net
943 networks showing relationships between isolates identified on the basis of the full set of
944 SNPs without missing data. The groups revealed are referred to as lineage C1, C2, and
945 C3. (B) Scatterplot from discriminant analysis of principal components (DAPC). (C)
946 Individual ancestry coefficients estimated using sNMF. Each isolate is represented by a
947 thick vertical line in the most probably number of groups ($K=3$), and bar colors represent
948 each lineage. Asterisks indicate admixture isolates.

949

950 **Figure 2.** Box plots of the average populations pairwise nucleotide changes per site in
951 100-kb windows within (nucleotide diversity (π)) and between lineages (d_{xy}).

952

953 **Figure 3.** Linkage disequilibrium (LD) decay plots of three genetic lineages of
954 *Colletotrichum truncatum* (C1, C2, and C3). LD measured was calculated for all pairs of
955 SNPs less than 300 kb distance apart.

956

957 **Figure 4.** Scan for signature of genetic exchanges between *C. truncatum* lineages. The
958 isolates LFN0297 (lineage C1), LFN0318 (lineage C2), and LFN0308 (lineage C3) were
959 used as representative of each lineage. The sequence similarity along contig 66 among
960 isolates visualized through RBG color triangular in the software HYBRIDCHECK. Areas
961 where two sequences have the same color (yellow, purple or turquoise) are indicative of
962 two lineages sharing the same polymorphisms. The bottom panel shows the linear plot
963 of the proportion of SNPs shared between the three pairwise comparisons.

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965 **Figure 5.** The best support demographic model (isolation with asymmetrical
966 migration) using the three-dimensional site frequency spectrum (3D-SFS) between
967 *Colletotrichum truncatum* lineages (C1, C2, and C3). The following parameters were
968 estimated: nu1: size of C1 after split; nu2: size of C2 after split; nu3: size of C3 after
969 split; m12: asymmetrical migration between C1 and C2; m21: asymmetrical migration
970 between C2 and C1; m23: asymmetrical migration between C2 and C3; m32:

971 asymmetrical migration between C3 and C2.; m13: asymmetrical migration between C1
972 and C3; m31: asymmetrical migration between C3 and C1; T1: scaled time between the
973 split and the size change (in units of $2 \cdot N_a$ generations). **Thick arrows indicate higher**
974 **migration rates.**

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976 **Supplemental figure captions**

977 **Figure S1.** Visual representation of the 3D demographic models fitted.

978

979 **Figure S2.** (A) Bayesian information criteria (BIC) indicating the most probable number
980 of genetic groups by discriminant analysis of principal components analysis (DAPC). (B)
981 Cross-entropy as a function of the number of clusters K modeled in sNMF analyses of
982 population subdivision.

983

984 **Figure S3.** Circos plot showing genome-wide diversity of the ten largest contigs of
985 *Colletotrichum truncatum* genomes. Nucleotide diversity (π) per SNP and Tajima's D
986 were calculated in sliding windows of 10 kb.

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988 **Figure S4.** Probabilistic chromosome painting bar plots of the ten largest contigs among
989 three *Colletotrichum truncatum* lineages (C1, C2, and C3).

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991 **Figure S5.** Variation in sequence similarity between *Colletotrichum truncatum* lineages
992 (C1, C2, and C3) on the ten largest contigs. The isolates LFN0297 (lineage C1),
993 LFN0318 (lineage C2), and LFN0308 (lineage C3) were used as representative of each
994 lineage and visualized as mosaic-like genome structure through RBG color triangular in
995 the software HYBRIDCHECK.

996

997 **Figure S6.** Demographic models tested using the three-dimensional site frequency
998 spectrum (3D-SFS) between *Colletotrichum truncatum* lineages (C1, C2, and C3). Data,
999 models and residuals are presented using a heatmap. The best model fitted is in bold. The
1000 following parameters were estimated: nu1: size of C1 after split; nu2: size of C2 after
1001 split; nu3: size of C3 after split; m12: asymmetrical migration between C1 and C2; m21:
1002 asymmetrical migration between C2 and C1; m23: asymmetrical migration between C2
1003 and C3; m32: asymmetrical migration between C3 and C2.; m13: asymmetrical migration
1004 between C1 and C3; m31: asymmetrical migration between C3 and C1; T1: scaled time

1005 between the split and the size change (in units of $2*N_a$ generations). LL = log likelihood
1006 of the model. AIC: Akaike information criterion.

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1037 **Table 1.** *Colletotrichum truncatum* isolates used in this study

Cluster	Isolate	State, city	Decimal coordinates	Cultivar	Sampling year
C2	LFN0217	Mato Grosso, Sinop	-11.923, -55.616	Nidera 7709	2017
C3	LFN0225	Mato Grosso, Sinop	-11.923, -55.616	Nidera 7709	2017
C2	LFN0248	Mato Grosso, Sinop	-11.982, -55.507	Nidera 7709	2017
C1	LFN0262	Mato Grosso, Sinop	-11.982, -55.507	Nidera 7709	2017
C3	LFN0268	Mato Grosso, Sinop	-11.982, -55.507	Nidera 7709	2017
C3	LFN0150	Mato Grosso, Lucas do Rio Verde	-13.169, -56.068	Y-70 Pioneer	2017
C1	LFN0169	Mato Grosso, Sinop	-13.312, -56.042	Monsoy 8766	2016
C1	LFN0185	Mato Grosso, Lucas do Rio Verde	-13.410, -56.067	Monsoy 8372	2017
C2	LFN0205	Mato Grosso, Lucas do Rio Verde	-13.410, -56.067	Monsoy 8372	2017
C2	LFN0288	Goiás, Montividiu	-17.411, -50.950	Monsoy 7739	2017
C3	LFN0291	Goiás, Montividiu	-17.411, -50.950	Monsoy 7739	2017
C1	LFN0297	Goiás, Montividiu	-17.411, -50.950	Monsoy 7739	2017
C3	LFN0308	Goiás, Rio Verde	-17.460, -51.119	Macroseed PP7200	2017
C1	LFN0309	Goiás, Rio Verde	-17.460, -51.119	Macroseed PP7200	2017
C2	LFN0318	Goiás, Rio Verde	-17.460, -51.119	Macroseed PP7200	2017
C1	LFN0346	Goiás, Rio Verde	-17.764, -51.035	Nidera 5909	2017
C2	LFN0349	Goiás, Rio Verde	-17.764, -51.035	Nidera 5909	2017
C1	LFN0360	Goiás, Rio Verde	-17.764, -51.035	Nidera 5909	2017

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1051 **Table 2.** Summary of genomic diversity within *Colletotrichum truncatum* lineages in
 1052 nonoverlapping 100kb windows

Lineage	N ^a	π^b	D ^c
C1	7	0.0002 (0.0003) *	0.04 (1.118)

C2	6	0.0062 (0.0041)	-0.380 (0.858)
C3	5	0.0113 (0.0056)	-0.180 (0.623)

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^a Sample size
^b Nucleotide diversity per base pair, for all contigs larger than 10kb
^c Tajima's neutrality statistic, for all contigs larger than 10kb
* Standard deviation

1081 **Table 3.** Percentage of genetic introgression between lineages of *Colletotrichum*
1082 *truncatum*

Pairwise lineages			
Contig	C1 x C2	C1 x C3	C2 x C3
tig55	2.70	13.56	11.17
tig66	1.88	1.31	68.43
tig97	-	-	67.74

tig164	-	36.00	-
tig209	1.99	20.08	6.74
tig284	-	0.22	19.95
tig332	4.38	-	12.55
tig526	0.03	1.19	78.54
tig70486	4.03	16.16	3.66
tig70488	0.22	2.23	11.87
total	2.43	12.75	28.72

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