Investigation of inter- and intraspecies variation through genome sequencing of Aspergillus section Nigri

Vesth, Tammi Camilla; Rasmussen, Jane Lind Nybo; Theobald, Sebastian; Frisvad, Jens Christian; Larsen, Thomas Ostenfeld; Nielsen, Kristian Fog; Hoof, Jakob Blæsbjerg; Brandl, Julian; Salamov, Asaf; Riley, Robert; Gladden, John M; Phatale, Pallavi; Nielsen, Morten Thrane; Lyhne, Ellen Kirstine; Kogle, Martin Engelhard; Strasser, Kimchi; McDonnell, Erin; Barry, Kerrie; Clum, Alicia; Chen, Cindy; LaButti, Kurt; Haridas, Sajeet; Nolan, Matt; Sandor, Laura; Kuo, Alan; Lipzen, Anna; Hainaut, Matthieu; Drula, Elodie; Tsang, Adrian; Magnuson, Jon K; Henrissat, Bernard; Wiebenga, Ad; Simmons, Blake A; Mäkelä, Miia R; de Vries, Ronald P; Grigoriev, Igor V; Mortensen, Uffe Hasbro; Baker, Scott E; Andersen, Mikael Rørdam

Published in:
Nature Genetics

Link to article, DOI:
10.1038/s41588-018-0246-1

Publication date:
2018

Document Version
Publisher's PDF, also known as Version of record

Link back to DTU Orbit

Citation (APA):
Investigation of inter- and intraspecies variation through genome sequencing of *Aspergillus* section *Nigri*


*Aspergillus* section *Nigri* comprises filamentous fungi relevant to biomedicine, bioenergy, health, and biotechnology. To learn more about what genetically sets these species apart, as well as about potential applications in biotechnology and biomedicine, we sequenced 23 genomes de novo, forming a full genome compendium for the section (26 species), as well as 6 *Aspergillus niger* isolates. This allowed us to quantify both inter- and intraspecies genomic variation. We further predicted 17,903 carbohydrate-active enzymes and 2,717 secondary metabolite gene clusters, which we condensed into 455 distinct families corresponding to compound classes, 49% of which are only found in single species. We performed metabolomics and genetic engineering to correlate genotypes to phenotypes, as demonstrated for the metabolite aurasperone, and by heterologous transfer of citrate production to *Aspergillus nidulans*. Experimental and computational analyses showed that both secondary metabolism and regulation are key factors that are significant in the delineation of *Aspergillus* species.

Species in the genus *Aspergillus* are of broad interest to medical, applied, and basic research. Members of *Aspergillus* section *Nigri* (‘black aspergilli’) are prolific producers of native and heterologous proteins, organic acids (in particular citric acid), and secondary metabolites (including biopharmaceuticals and mycotoxins like ochratoxin A). Furthermore, the section members are generally very efficient producers of extracellular enzymes; they are the production organisms for 49 out of 260 industrial enzymes, 49 out of 260 industrial enzymes. Among the most important of these, in addition to *A. niger*, are *A. tubingensis*, *A. aculeatus*, and *A. luchuensis* (previously *A. acidus*, *A. kawachii*, and *A. awamori* respectively).

Members of *Aspergillus* section *Nigri* are also known as degradative species of foods and feeds, and some isolates produce the potent mycotoxins ochratoxin A and fumonisins. In addition, some species in this section have been proposed to be pathogenic to humans and other animals. It is thus of interest to further examine section *Nigri* for industrial exploitation, as well as prevention of food spoilage, toxin production, and pathogenicity caused by these fungi.

A combined phylogenetic and phenotypic approach has shown that section *Nigri* contains at least 27 species. Recent results have shown that the section contains species with high diversity and may consist of two separate clades: the biseriate species and the uniseriate species, which show differences in sexual states, sclerotium formation, and secondary metabolite production. In the section, only six species have had their genome sequenced: *A. niger*, *A. luchuensis*, *A. carbonarius*, *A. aculeatus*, *A. tubingensis*, and *A. brasiliensis*.

This section, with its combination of species richness and fungal species with a diverse impact on humanity, is thus particularly interesting for studying the diversification of fungi into species. In this study, we have de novo-sequenced the genomes of 20 species of section *Nigri*, thus completing a genome compendium of 26 described species in the section. Further, we have genome-sequenced three...
Additional A. niger isolates (including two previously described as species A. lactoceffacus and A. phoenicis), which in combination with the other analyses allows for inter- and intraspecies comparison of 32 isolates. The development of algorithms for comparative genomics, combined with experimental analysis of the species, allows us to track genetic diversity across genomes, from the protein level, over the evolution of biosynthetic gene clusters, to the groups of genes that define clades or individual species. The high resolution in genome sequences allows us to characterize both species diversification and variation within species.

Results
New genomes show high genetic diversity of section Nigri. We present 23 whole-genome draft sequences: 20 of section Nigri species previously unsequenced and 3 additional A. niger genomes for assessment of intraspecies diversity. All genomes were sequenced, assembled, and annotated using the Joint Genome Institute (JGI) fungal genome pipeline (Supplementary Table 1; genomes were sequenced by either Illumina or Pacific Biosciences sequencing). Figure 1 shows a phylogenetic tree as well as gene richness, number of scaffolds, and functional annotation (InterPro). The tree supports previous proposals that A. lactoceffacus and A. phoenicis are synonyms of A. niger.

In comparing key statistics of the genomes, we found that some traits are quite similar and others surprisingly variable. Many of the investigated species have around the average number of genes (11,900), but there is considerable variation from the smallest number of predicted genes (10,066) to the largest (13,687). The smallest number of predicted genes in section Nigri is found in A. saccharolyticus, which supports the previous observation that this species is quite atypical in section Nigri.

We further evaluated the annotation of the 23 genome sequences we generated. The percentage of complete genes (including a start and stop codon) is in the range of 94–98%, and 67% of the proteins could be assigned one or more InterPro domains. The number of proteins with homologs in Swiss-Prot (91% of proteins have homologs within section Nigri). On average, 70% of the proteins had sequence homology to proteins in the Swiss-Prot database.

The pan- and core-genome shows genome flexibility. Given the genetic diversity in section Nigri, we were interested in examining the extent of genome diversification. For this analysis, we focused on three conceptual groups of genes:

1. The pan-genome: all genes present in one or more species.
2. The core-genome: genes present in all included species.
including paralogs. This set is expected to encode cellular functions needed for all species. 

(3) Species-unique genes: genes found in only one species in our analysis, with or without paralogs. Included in these, we would expect to find genes involved in environmental adaptation. This group can also include annotation errors.

We first identified orthologs and paralogs with a BLASTp-based pipeline using reciprocal hits according to cut-offs specifically selected here for the Aspergillus genus (Methods). Groups of homologous proteins are referred to as families. Figure 2a–c shows the overall genetic diversity between 38 fungal strains (32 species) from closely related genera (Fig. 2a), within the Aspergillus genus (36 of the 38 strains; Fig. 2b), and from section Nigri (32 of the 38 strains; Fig. 2c).

The Aspergillus genus pan-genome comprises 433,116 genes across the 36 Aspergillus genomes, and from this, 62,996 gene families were constructed. Of those families, 6% are found in all genomes (3,769 core families), while 9% are genes without orthologs in the other genomes (40,424 unique genes; 39,929 unique families) (Fig. 2b). We also found evidence of gene loss, duplication, and potential gene transfers between species of this section, as 23% of the pan-gene families are not present in groups of species fitting the phylogenetic tree (Supplementary Table 2). This is consistent with previous work reporting extensive horizontal gene transfer in Aspergillus39.

We further performed an analysis defining the number of core-gene families in section Nigri and in all sub-clades thereof (Fig. 2d). The core-genome of section Nigri is 32% larger than that of the genus (4,983 families relative to 3,769; Fig. 2b,c). Conversely, 9% are unique to a specific species (32,378 unique genes in 32,036 families; Fig. 2c). The fraction of genes unique to a species is similar within the section and across the genus, meaning that adding a new section Nigri genome adds as many new genes as adding a more distantly related Aspergillus (within the analyzed group of species). This is rather interesting and shows a generally high genetic diversity of
The section *Nigri* core genome contains carbohydrate-active enzymes and secondary metabolism gene clusters. To associate biological functions to the pan-, core-, and unique genomes, and genes exclusive to only members of the black aspergilli, we employed the InterPro database. An examination of the core-genome of 38 fungal genomes (Fig. 2a, Supplementary Table 1) revealed that only 4.5% of the genes lack InterPro domains (Supplementary Table 3a), indicating—as would be expected—that the core-genes across closely related fungal genera include generally known and conserved functions. For the pan-genomes of the 36 *Aspergillus* species compared with the section *Nigri* genomes, the percentages of unknown function are similar (32% compared with 33%; Supplementary Tables 3d and 4a and Supplementary Fig. 2), as are the corresponding percentages for the core-genes (14% compared with 17%, Fig. 2c; Supplementary Tables 3d and 4a). General functions like transporters, regulators, organelle-specific proteins, primary metabolism, and structural domains were found as core features across all 36 aspergilli (Supplementary Table 3f), which supports the general validity of the method.

We expected the section *Nigri* core-genome (gene families found in all the species in section *Nigri* but not in any other aspergilli examined) to contain *Nigri* signature genes, and we found this to be the case. These 1,214 gene families contain 580 InterPro domains, a many genes involved in the saprotrophic lifestyle and secondary metabolism (Supplementary Table 5). It is hypothesized that these genes are defining for the section compared with other aspergilli and will encode functions related to the phenotypes of species in this section.

Unique secondary metabolism genes in *Aspergillus* species. The genetic diversity seen in section *Nigri* led us to investigate whether the unique genes for each species show common trends in function. While these genes by definition do not have homologs in other species investigated in this work, we can predict general functions using InterPro domains. Unique genes of species in section *Nigri* matched 1,334 different InterPro domains (Supplementary Table 6a–c). Within the unique genes, we searched the list of InterPro domains in all sets of genes unique to individual section *Nigri* species (excluding the six *A. niger* isolates, to remove intraspecies redundancy). Surprisingly, we identified only ten domains that were found in nearly all *Nigri* species (25–26 species). Notably, nine of those are related to functions involved in secondary metabolism, gene regulation (transcription factors), or protein regulation (protein kinases) (Supplementary Table 7). Finding these functions in nearly all sets of species-specific genes suggests that secondary metabolite production and regulatory proteins are commonly identified as the species-‘unique’ genes and are therefore critical differentiates for fungal species at the genetic level.

Intra- and interspecies genetic variations are similar. We were interested in comparing the diversity between isolates of the same species to the diversity among species in the same clade. We thus compared six *A. niger* isolates to the eight closely related species in the *A. tubingensis* clade (Fig. 2d). The *A. niger* isolates have a high
degree of genetic homogeneity, as 80% of the A. niger pan-genome is conserved across the six isolates and only 6% is unique to any of the isolates (Supplementary Fig. 3a). The same scale is seen in the A. tubingensis clade (77% shared pan-genome, 7% unique; Supplementary Fig. 3b). Moreover, the percentage of genes with predicted functions seen here (Supplementary Table 11). Considering the relative uniformity of the CAZyme content (Fig. 4), no correlation between genome content and growth on plant biomass-related carbon sources (Supplementary Table 11) and performed growth profiling on plant biomass-related carbon sources (Supplementary Fig. 6). Growth on d-glucose was used to evaluate relative growth, showing variation between species.

In a previous study, enzyme levels were measured in several black aspergilli, and significant differences were found. However, differences in enzyme levels do not reflect the copy number differences seen here (Supplementary Table 11). Considering the relative uniformity of the CAZyme content (Fig. 4), no correlation between genome content and growth on plant biomass-related carbon sources (Supplementary Fig. 6) was observed for the black aspergilli, suggesting that the differences in capability for plant biomass degradation reflect gene expression levels in the individual fungus. This confirms a proteome study of less-related aspergilli, in which the different response to plant biomass appeared to be mainly at the regulatory level. The data suggest that this is the case for section Nigri: species-specific phenotypes are driven not generally by CAZyme content in closely related species, but by regulation.

Secondary metabolism in section Nigri contains 455 families. Secondary metabolism is thought to be a component of chemical defense, virulence, toxicity, mineral uptake, and communication in fungi and has a wide range of potential medical applications. As we had identified it to be commonly unique to individual species, we examined the exometabolite diversity of 37 Aspergillus and Penicillium species according to predictions of secondary
metabolism gene clusters (SMGCs) as well as chemical profiles of the species of section Nigri on multiple substrates.

We identified 2,717 SMGCs in the 37 genomes. This is an even higher number of SMGCs per species than a previous study found in 24 Penicillium genomes. We were further interested in quantifying the actual diversity of the SMGCs in section Nigri and in analyzing presence patterns of SMGCs across species. We therefore defined SMGC 'families' as genetically similar SMGCs across genomes (Methods). Each SMGC family is expected to produce the same or similar compounds. This clustering resulted in the definition of 455 SMGC families across the 37 genomes (Supplementary Fig. 7), indicating the potential production of 455 different chemical families. Most families (82%) are found in fewer than 10 organisms, 49% contain only one gene cluster (Supplementary Fig. 8 shows examples). On average there are 8.75 unique clusters per species, despite the close phylogenetic distance of the section.

Phylogenetic examination shows dynamic content of SMGCs. To reveal more about how SMGCs evolve and differentiate between species, each of the 455 SMGC families was characterized by the type of backbone enzyme and analyzed according to the phylogeny (Fig. 5a, b). Only five out of all SMGCs were present in all analyzed species, including clusters for the non-ribosomal peptide synthetase (NRPS)-derived siderophore ferrichrome, the circular NRP fungisporin/ nidulanin A, and pigment (YWAI) synthesis. Two shared SMGC families were false predictions, namely two fatty acid synthases.

Examining the dynamics of the families, only 32% and 19% of SMGCs found in two or three organisms, respectively, follow the whole-genome phylogeny and suggest intragenus horizontal gene transfer or SMGC loss to be relatively common. As an example, an SMGC is found in five distantly related species (Supplementary Table 1). The cluster is found in all A. niger isolates as well as in A. tubingensis, A. homomorphus, A. gibbosa, A. homomorphus, and A. tubingensis.

As seen in Fig. 5a, the presence of unique SMGC families at every major branch point in the phylogenetic tree supports that SMGCs are a part of what sets the species apart: all biseriates share a previously undescribed polyketide synthase (PKS) and an NRPS-like protein, the A. carbonarius clade a terpene cyclase, and the remaining biseriates share another PKS and another NRPS-like protein. Furthermore, the A. niger complex and A. tubigenis clade each share unique PKS genes. Uniseriates share four unique previously undescribed SMGC families (Fig. 5a). Examinations of individual species reveal that every single section Nigri species has a unique combination of SMGCs (Fig. 3b). Furthermore, nearly all
species genomes (with the exception of A. tubingenesis, A. niger, A. brasiliensis, and A. vandenensis) encode one or more unique SMGCs. These patterns show the existence of high diversity of SMGCs between species and of a homogenous set of SMGCs within isolates from the same species.

Correlating secondary metabolisms with SMGC families links gene to function. As a further application of the constructed SMGC families, we hypothesized that we can correlate SMGC families to classes of compounds. We performed extensive exometabolome analysis of 27 of the sequenced strains and identified 35 compound families (Fig. 5c and Supplementary Table 12).

The most abundant group was naphtha-γ-pyrones, of which aurasperone B29 was identified in 14 of the isolates. We compared the presence patterns of SMGC families with the compound class (Fig. 5c) and combined it with a knowledge-based filtering of InterPro domains leaving one hit (Methods and Supplementary Fig. 8d). The candidate SMGC family is a nine-gene cluster found in 18 genomes—including the 14 where we detected the compound—and it contains all activities needed to synthesize aurasperone. In support of this identification, an SMGC for a closely related compound, aurofusarin, has been experimentally verified in Fusarium graminearum29. The aurasperone cluster shares six genes (one of which is a duplication) with the aurofusarin cluster. This finding supports the assignment of this family of SMGCs to the production of aurasperone B and conceptually justifies this approach for efficient linking of clusters to compounds. We see this correlation approach as highly useful for future elucidation of fungal metabolites.

Discussion
We have sequenced the genomes of a whole section of filamentous fungi, and a diverse set of A. niger isolates, and found that the species are highly diverse in some traits, in particular secondary metabolism and to a lesser extent regulatory proteins, and homogeneous in others, such as glycolytic metabolism and CAZymes. The presented data furthermore provide an extensive compendium of 24 new genes, which adds substantial information on fungal genetic diversity. We further combined genome analysis with metabologe profiling and heterologous gene expression to identify the genetic basis of several phenotypes within primary and secondary metabolism.

Of particular interest was the finding that the species-specific genes in all species share functions within gene/protein regulation and secondary metabolism, showing that unique sets of these functions exist for all species in the investigated set.


Online content
Any methods, additional references, Nature Research reporting summaries, source data, statements of data availability and associated accession codes are available at https://doi.org/10.1038/s41588-018-0246-1.

Received: 12 February 2018; Accepted: 23 August 2018; Published online: 22 October 2018

References

**Acknowledgements**

T.C.V., J.L.N., and M.R.A. acknowledge funding from The Villing Foundation/ Villing Fonden, grant VKR023437. J.B. and M.R.A. acknowledge funding from the Novo Nordisk Foundation, grant NNF13OC0004831. M.R.A. and T.C.V. acknowledge funding from the Novo Nordisk Foundation, grant NNF13OC005201. Work performed at the US Department of Energy (DOE) Joint BioEnergy Institute is supported by the US DOE, Office of Science, Office of Biological and Environmental Research, through Contract No. DE-AC02-05CH11231 between Lawrence Berkeley National Laboratory and the US DOE. The work conducted by the US-DOE Joint Genome Institute, a DOE Office of Science User Facility, is supported by the Office of Science of the US DOE under Contract No. DE-AC02-05CH11231 between Lawrence Berkeley National Laboratory and the US DOE. S. Mondo is acknowledged for GenBank submission of 20 of the genomes. I. Kjerbolling is acknowledged for critical and constructive feedback on the manuscript and figures.

**Author contributions**

J.L.N. and S.T. analyzed data, designed and generated algorithms, contributed to design of research, and wrote parts of the manuscript. I.C.F. contributed to design of research, contributed analytical tools and data for species selection and verification, wrote parts of the manuscript, and analyzed data. T.O.L. and K.E.N. generated data on secondary metabolism, analyzed data, and wrote parts of the manuscript. J.B. and M.T.N. engineered strains, analyzed data, and wrote parts of the manuscript. L.B. analyzed data on primary metabolism and wrote parts of the manuscript. A.S., R.B., A.K., and S.H. annotated genomes and analyzed data. J.M.G. and J.K.M. contributed to design of research and contributed analytical tools. P.P. contributed analytical tools. E.K.L. and M.E.K. contributed to design of research, developed methods, conducted experiments, wrote parts of the manuscript, and analyzed data. C.C. and L.S. sequenced RNA and DNA. M.N., A.L., K.L., and A.C. assembled the genomes. M.H., E.D., and B.H. contributed analytical tools and analyzed CAZyme data. A.W. performed part of the experiments. M.R.M. analyzed data and wrote part of the manuscript. R.P.d.V. analyzed data and wrote part of the manuscript. A.T., K.S., and E.M. contributed to design of research, contributed analytical tools and data, and analyzed data. K.B. and L.V.G. coordinated the DNA and RNA sequencing and annotation. B.A.S. contributed to design of research. U.H.M. designed parts of the experiments and developed methods. S.E.B. conceived the overall project, analyzed data, contributed to design of research, and contributed to writing and editing the manuscript. T.C.V. and M.R.A. conceived the overall project, analyzed data, contributed to design of research, designed algorithms, wrote parts of the manuscript, and coordinated the project. All authors commented on the manuscript.

**Competing interests**

The authors declare no competing interests.

**Additional information**

Supplementary information is available for this paper at https://doi.org/10.1038/s41588-018-0246-1.

Reprints and permissions information is available at www.nature.com/reprints.

Correspondence and requests for materials should be addressed to S.E.B. or M.R.A.

Publisher’s note: Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.

---

**Open Access** This article is licensed under a Creative Commons Attribution 4.0 International License, which permits use, sharing, adaptation, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons license, and indicate if changes were made. The images or other third party material in this article are included in the article’s Creative Commons license, unless indicated otherwise in a credit line to the material. If material is not included in the article’s Creative Commons license and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this license, visit http://creativecommons.org/licenses/by/4.0/.

© This is a U.S. government work and not under copyright protection in the U.S.; foreign copyright protection may apply 2018
Methods
Fungal strains. Unless otherwise noted, the species examined were taken from the IBT Culture Collection of Fungi at DTU. Strains employed in this study are denoted in Supplementary Table 1.

Purification of DNA and RNA. For all sequences generated for this study (Supplementary Table 1), spores were defrosted from storage at −80 °C and inoculated onto solid CYA medium. Fresh spores were harvested after 7–10 d and suspended in a 0.1% Tween solution. Spores were stored in solution at 5 °C for up to 3 weeks. Biomass for all fungal strains was obtained from shake flasks containing 200 ml of complex medium, either CYA, MEAox, or CY20 depending on the strain (see Supplementary Table 1) cultured for 5–10 d at 30 °C. Biomass was isolated by filtering through Miracloth (Millipore, 47.85x1–1), freeze dried, and stored at 80 °C.

Purification of DNA and RNA isolation was performed using a modified version of the standard protocol (ref. 1 and below) and checked for quality and concentration using a NanoDrop (Thermo Scientific). RNA isolation was performed using the Qiagen RNeasy Plant Mini Kit according to the manufacturer’s instructions.

A sample of frozen biomass was subsequently used for RNA purification. First, hyphae were transferred to a 2 ml microtube together with a 5 mm steel bead (Qiagen), placed in liquid nitrogen, then lysed using the Qiagen TissueLyser LT at 45 Hz for 50 s. Then the Qiagen RNeasy Mini Plus Kit was used to isolate RNA. RLT Plus buffer (with 2-mercaptoethanol) was added to the samples, vortexed, and spun down. The lysate was then used in step 4 in the instructions provided by the manufacturer, and the protocol was followed from this step. For genomic DNA, a protocol based on Fultton et al.17 was used (See Supplementary Note).

DNA and RNA sequencing and assembly. All genomes in this study, except for those of A. heteromorphus, A. eucalypticola, and A. sclerotiorum, and all transcriptomes were sequenced with Illumina. The genomes of A. heteromorphus, A. eucalypticola, and A. sclerotiorum were sequenced with PacBio.

For all genomic Illumina libraries, 100- ng of DNA was sheared to 270 bp fragments using the Covaris LE220 (Covaris) and size selected using SPRI beads (Beckman Coulter). The fragments were treated with end-repair and A-tailing and ligated to Illumina-compatible adapters (IDT) using the KAPA- Illumina library creation kit (KAPA Biosystems).

For transcriptomes, stranded complementary DNA libraries were generated using the Illumina TruSeq Stranded Total RNA LT Sample Prep Kit. Messenger RNA (mRNA) was purified from 1 µg of total RNA using magnetic beads containing poly(T) oligos. mRNA was fragmented using divalent cations and high temperature. The fragmented RNA was reverse transcribed using random hexamers and SSIII (Invitrogen) followed by second-strand synthesis. The fragmented complementary DNA was treated with end-pair, A-tailing, adapter ligation, and 16 cycles of PCR.

The prepared libraries were quantified using KAPA Biosystems’ next-generation sequencing library quantitative PCR kit and run on a Roche LightCycler 480 real-time PCR instrument. The quantified libraries were then multiplexed with other libraries, and library pools were prepared for sequencing on the Illumina MiSeq sequencing platform using a TruSeq paired-end cluster kit, v3, and Illumina’s cbot instrument to generate clustered flow cells for sequencing.

Sequencing of the flow cells was performed on the Illumina HiSeq2000 sequencer using a TruSeq SBS sequencing kit, v3, following a 2 x 150 indexed run recipe. After sequencing, the genomic FASTQ files were quality control-filtered and the reads of reads longer than 100 bp were used for assembly.

Some samples were analyzed in ESI -full-scan mode, scanning m/z 100–1,700. Data were analyzed by aggressive dereplication19 using lists of compounds considered to be from black aspergilli only (~350); a list with all Aspergillus compounds (~2,450); and a list of 1,600 reference standards, of which 500 are known to come from Aspergillus. Unknown peaks were matched against Antibase20 and dereplicated using several mass isotopomers, patterned adducts, patterns, and ultraviolet-visible data21.

Method 3. UHPLC–DAD–HRMS was conducted on an Agilent Infinity 1290 UHPLC system coupled to an Agilent 6550 QTOF MS. Separation was obtained on an Agilent Poroshell 120 phenyl-hexyl column (2.1 x 25 mm, 2.7 µm) using a linear gradient of water and acetonitrile (both buffered with 20 mM formic acid), progressing from 10% acetonitrile and increasing to 100% (both containing 50 ppm trifluoroacetic acid) over 8 min, then using 100% acetonitrile for 2 min. The column temperature was 60 °C, the flow rate 0.8 ml min−1, and the injection volume was 1 µl. The ultraviolet spectra 200–640 nm were matched against our internal database18.

Analysis of secondary metabolism. Cultivation for secondary metabolite analysis. Fungal strains were cultivated as three-point cultures on CYA, CYA5, and YES media for 7 d in the dark at 25 °C. Three 6 mm inner diameter plugs taken across the cultures were then extracted using an (3:2:1) ethylacetate–dichloromethane–methanol mixture and dissolved in methanol22.

Extraction of fungal metabolites. Fungal metabolite extracts were prepared using one of the three following methods: (1) chloroform–methanol–ethylacetate extraction, (2) micro-extraction using methanol–dichloromethane–ethylacetate, or (3) 75% methanol extraction.

Chemical analysis of secondary metabolites. All chemical analyses were done by reversed-phase ultrahigh-performance liquid chromatography (UHPLC) coupled to ultraviolet–visible diode array detection (DAD) combined with either fluorescence detection (FLD) or high-resolution mass spectrometry (HRMS). Three different methods were used:

Method 1. Pure UHPLC–DAD–FLD was performed using a rapid-separation liquid chromatography (RSLC) UltiMate 3000 system (Dionex) linked to an 1100 Series FLD (Agilent). The system was equipped with an Agilent Poroshell phenyl-hexyl column (150 x 2.1 mm, 6.2 µm) and was run using a linear gradient of water–acetonitrile starting at 10% acetonitrile and increasing to 100% (both containing 50 ppm trifluoroacetic acid) over 8 min, then using 100% acetonitrile for 2 min. The column temperature was 60 °C, the flow rate 0.8 ml min −1, and the injection volume was 1 µl. The ultraviolet spectra 200–640 nm were matched against our internal database18.

Method 2. UHPLC–DAD–HRMS was conducted on a Dionex RSLC UltiMate system linked to a mXis high-definition quadrupole–time–of–flight mass spectrometer (Q–TOF MS) (Bruker Daltonics). Separation was done on a Kinetex C18 column (100 x 2.1 mm, 6.2 µm), with a linear gradient consisting of water and acetonitrile (both buffered with 20 mM formic acid), starting at 10% acetonitrile and increasing to 100% over 10 min, where it was held for 2 min and returned (0.4 ml min−1, 40 °C). Injection volume, depending on sample concentration, typically varied between 0.1 and 1 µl. Some samples were analyzed in electrospray ionization (ESI) and some in ESI–full-scan mode, scanning m/z 100–1,250. Data were analyzed by aggressive dereplication19 using lists of compounds considered to be from black aspergilli only (~350); a list with all Aspergillus compounds (~2,450); and a list of 1,600 reference standards, of which 500 are known to come from Aspergillus. Unknown peaks were matched against Antibase20 and dereplicated using several mass isotopomers, patterned adducts, patterns, and ultraviolet-visible data21.
constructed on the basis of best bidirectional hits. Two hundred groups with a member from each species were selected, and the sequences of each organism were concatenated into one long protein sequence. Concatenated sequences were aligned using MAFFT (thread 16), and well-aligned regions were extracted using gblocks (–t g–4–5–6–8–20). Trees were then constructed using multithreaded RAXML, the PROTGAMMAWAG model, and 100 bootstrap replicates.

**Prediction of SMGCs.** For the prediction of SMGCs, we developed a command-line Python script roughly following the SMURF algorithm:

```python
As input, the program takes genomic coordinates and the annotated PFAM domains of the predicted genes. Based on the multifast domain PFAM composition of identified 'backbone' genes, it can predict seven types of secondary metabolite clusters: (1) polyketide synthases (PKSs), (2) PKS-like, (3) non-ribosomal peptide-synthetases (NRPSs), (4) NRPS-like, (5) hybrid PKS-NRPS, (6) prenyltransferases (DMATS), and (7) terpene cyclases (TCs). Besides backbone genes, PFAM domains, which are enriched in experimentally identified secondary metabolite clusters (secondary metabolite-specific PFAMs), were used in determining the borders of gene clusters. The maximum allowed size of intergenic regions in a cluster was set to 3 kb, and each predicted cluster was allowed to have up to 6 genes without secondary metabolite-specific domains.

**Prediction of secreted proteases.** Secretome prediction was done using an in-house adaptation of SignalP69. Presence/absence of an orthologous gene to a member in a gene cluster was based on a bidirectional best hit, with e < 1 × 10\(^{-10}\) and coverage of >90%. Presence/absence of a full gene cluster was based on the occurrence of the majority of the predicted members in a gene cluster, including the backbone synthetase in another species. Prediction of secreted proteases. Secretome prediction was done using an in-house adaptation of SignalP69.

**Gene-compound assignment.** Identification of conserved or highly similar fungal gene clusters was performed on the basis of the gene cluster predictions above. The genomes were compared using the BLASTp function from the BLAST+ suite70. As input, the program takes genomic coordinates and the annotated PFAM domains of the predicted genes. Based on the multifast domain PFAM composition of identified 'backbone' genes, it can predict seven types of secondary metabolite clusters: (1) polyketide synthases (PKSs), (2) PKS-like, (3) non-ribosomal peptide-synthetases (NRPSs), (4) NRPS-like, (5) hybrid PKS-NRPS, (6) prenyltransferases (DMATS), and (7) terpene cyclases (TCs). Besides backbone genes, PFAM domains, which are enriched in experimentally identified secondary metabolite clusters (secondary metabolite-specific PFAMs), were used in determining the borders of gene clusters. The maximum allowed size of intergenic regions in a cluster was set to 3 kb, and each predicted cluster was allowed to have up to 6 genes without secondary metabolite-specific domains.

**Detection of encoded CAZymes.** Each Aspergillus protein model was compared using BLASTp to proteins listed in the Carbohydrate-Active Enzymes database (CAZY)84. Models with over 50% identity over the entire length of an entry in CAZY were directly assigned to the same family (or subfamily when relevant). Proteins with less than 50% identity to a protein in CAZY were all manually inspected, and conserved features, such as the catalytic residues, were searched whenever known. Because 30% sequence identity results in widely different e-values (from non-significant to highly significant), for CAZY family assignments, we examined sequence conservation (percentage identity over CAZY domain length). Sequence alignments with isolated functional domains were performed in the case of multimodular CAZymes. The same methods were used for Penicillium chrysogenum and Neurospora crassa.

**Mapping of genes shared by groups of species.** All predicted sets of protein sequences for the 38 genomes analyzed were aligned using the BLASTp function from the BLAST+ suite version 2.2.27 (e-value cut-off ≤ 1 × 10\(^{-10}\)). These 1,444 whole-genome BLAST tables were analyzed to identify bidirectional hits in all pairwise comparisons. Using custom Python scripts, homologs were identified within and across the genomes and grouped into sequence-similar families using single linkage, if they met the following criterion: The sum of the alignment coverage between the pairwise sequences was >10%, the alignments identity between the pairwise sequences was >50%, and the hit must be found in both of the species’ BLAST output (reciprocal hits). Singleton families were assigned a family having only one gene member. This allowed for identification of species-unique genes as well as genes shared by sections, clades, and sub-clades of species. All homologous were assigned functional and structural domains using InterPro version 48 and checked for annotation and sequencing errors by investigating scaffold location and sequence identity.

For the analysis of the pan- and core-genomes of a subset of 38 fungal species used in this study, the orthologous and paralogous families were subsetted to include only the species of interest. Therefore, the genes representing the core unique portion of the genomes will adjust relative to the accompanying species.

**Identification of SMGC families.** Our implementation of SMURF was run on genomic data from 37 Aspergillus strains. Proteins of the resulting SMGCs were compared with each other by alignment using BLASTp (BLAST+ suite version 2.2.27, e-value cut-off ≤ 1 × 10\(^{-10}\)). Subsequently, a score based on BLASTp identity and shared proteins was created to determine the similarity between gene clusters as depicted in the formula below. Using these scores, we created a weighted network of SMGC clusters and used a random walk community detection algorithm (R version 3.3.2, igraph_1.0.1+) to determine families of SMGC clusters. Finally, we ran another round of random walk clustering on the communities that contained more members than the species in the analysis (ptailoring/pbackbone = sum percentage BLAST alignment of tailoring/backbone enzymes, respectively; ntailoring/backbone = number of tailoring/backbone enzymes with significant hits, respectively; ptailoring/pbackbone = total number of tailoring/backbone enzymes):
References

Reporting Summary

Nature Research wishes to improve the reproducibility of the work that we publish. This form provides structure for consistency and transparency in reporting. For further information on Nature Research policies, see Authors & Referees and the Editorial Policy Checklist.

Statistical parameters

When statistical analyses are reported, confirm that the following items are present in the relevant location (e.g. figure legend, table legend, main text, or Methods section).

- None
- n/a
- Confirmed

☐ The exact sample size (n) for each experimental group/condition, given as a discrete number and unit of measurement
☐ An indication of whether measurements were taken from distinct samples or whether the same sample was measured repeatedly
☐ The statistical test(s) used AND whether they are one- or two-sided
☐ Only common tests should be described solely by name; describe more complex techniques in the Methods section.
☐ A description of all covariates tested
☐ A description of any assumptions or corrections, such as tests of normality and adjustment for multiple comparisons
☐ A full description of the statistics including central tendency (e.g. means) or other basic estimates (e.g. regression coefficient) AND variation (e.g. standard deviation) or associated estimates of uncertainty (e.g. confidence intervals)
☐ For null hypothesis testing, the test statistic (e.g. F, t, r) with confidence intervals, effect sizes, degrees of freedom and P value noted
☐ Give P values as exact values whenever suitable.
☐ For Bayesian analysis, information on the choice of priors and Markov chain Monte Carlo settings
☐ For hierarchical and complex designs, identification of the appropriate level for tests and full reporting of outcomes
☐ Estimates of effect sizes (e.g. Cohen’s d, Pearson’s r), indicating how they were calculated
☐ Clearly defined error bars
☐ State explicitly what error bars represent (e.g. SD, SE, CI)

Our web collection on statistics for biologists may be useful.

Software and code

Policy information about availability of computer code

- Data collection: No software was used for data collection
- Data analysis: Custom code was used, this is available through GitHub: https://github.com/RoerdamAndersenLab/

For manuscripts utilizing custom algorithms or software that are central to the research but not yet described in published literature, software must be made available to editors/reviewers upon request. We strongly encourage code deposition in a community repository (e.g. GitHub). See the Nature Research guidelines for submitting code & software for further information.

Data

Policy information about availability of data

All manuscripts must include a data availability statement. This statement should provide the following information, where applicable:

- Accession codes, unique identifiers, or web links for publicly available datasets
- A list of figures that have associated raw data
- A description of any restrictions on data availability

Genome assembly and annotations are available at the JGI fungal genome portal MycoCosm (6) (http://jgi.doe.gov/fungi) and have been deposited at DDBJ/EMBL/GenBank under the following accessions: A. aculeatinus (PSTE00000000), A. bruneoviolaceus (PSTC00000000), A. costaricaensis (PSTH00000000), A. ellipticus (PSSY00000000), A. eucalypticola (MSFU00000000), A. fijiensis (PSTG00000000), A. heteromorphus (MSPL00000000), A. homomorphus (PSTJ00000000), A. ibericus.
Field-specific reporting

Please select the best fit for your research. If you are not sure, read the appropriate sections before making your selection.

- Life sciences
- Behavioural & social sciences
- Ecological, evolutionary & environmental sciences

For a reference copy of the document with all sections, see nature.com/authors/policies/ReportingSummary-flat.pdf

Life sciences study design

All studies must disclose on these points even when the disclosure is negative.

<table>
<thead>
<tr>
<th>Sample size</th>
<th>N/A</th>
</tr>
</thead>
<tbody>
<tr>
<td>Data exclusions</td>
<td>N/A</td>
</tr>
<tr>
<td>Replication</td>
<td>Genomes were only sequenced once each, but this is in accordance with current best practice.</td>
</tr>
<tr>
<td>Randomization</td>
<td>N/A</td>
</tr>
<tr>
<td>Blinding</td>
<td>N/A</td>
</tr>
</tbody>
</table>

Reporting for specific materials, systems and methods

<table>
<thead>
<tr>
<th>Materials &amp; experimental systems</th>
<th>Methods</th>
</tr>
</thead>
<tbody>
<tr>
<td>n/a Involving the study</td>
<td>n/a Involving the study</td>
</tr>
<tr>
<td>√ × Unique biological materials</td>
<td>√ × ChIP-seq</td>
</tr>
<tr>
<td>√ × Antibodies</td>
<td>√ Flow cytometry</td>
</tr>
<tr>
<td>√ × Eukaryotic cell lines</td>
<td>√ MRI-based neuroimaging</td>
</tr>
<tr>
<td>√ × Palaeontology</td>
<td></td>
</tr>
<tr>
<td>√ × Animals and other organisms</td>
<td></td>
</tr>
<tr>
<td>× × Human research participants</td>
<td></td>
</tr>
</tbody>
</table>

Unique biological materials

Policy information about availability of materials

Obtaining unique materials Strains are available from the authors, from strain collections and/or from the original isolators of the material.