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

Tammi C. Vesth1, Jane L. Nybo2, Sebastian Theobald1, Jens C. Frisvad1, Thomas O. Larsen1, Kristian F. Nielsen1, Jakob H. Hoof1, Julian Brandl1, Asaf Salamov2, Robert Riley1,15, John M. Gladden2,3,4, Pallavi Phatale3,5, Morten T. Nielsen1, Ellen K. Lyhne1, Martin E. Kogle1, Kimchi Strasser6, Erin McDonnell6, Kerrie Barry7, Alicia Clum2, Cindy Chen2, Kurt LaButti2, Sajeet Haridas2, Matt Nolan2, Laura Sandor2, Alan Kuo2, Anna Lipzen2, Matthieu Hainaut7,8, Elodie Drula7,8, Adrian Tsang6, Jon K. Magnuson3,5, Bernard Henrissat7,8,9, Ad Wiebenga10, Blake A. Simmons6,3,11, Miia R. Mäkelä8,10,12, Ronald P. de Vries10, Igor V. Grigoriev2,13, Uffe H. Mortensen1, Scott E. Baker3,14* and Mikael R. Andersen1*.

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, applied1, and basic research4. Members of Aspergillus section Nigri (‘black aspergilli’) are prolific producers of native and heterologous proteins5–8, organic acids (in particular citric acid2,7,8), and secondary metabolites (including biopharmaceuticals and mycotoxins like ochratoxin A). Furthermore, the section members are generally very efficient producers of extracellular enzymes9,10; they are the production organisms for 49 out of 260 industrial mycotoxins like ochratoxin A16 and fumonisins17–19. In addition, 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 species1–25. 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 species26, which show differences in sexual states27, sclerotium formation28, and secondary metabolite production29. In the section, only six species have had their genome sequenced: A. niger2,4, A. luchuensis15–17, A. carbonarius31, A. aculeatus31, A. tubingensis31, and A. brasiliensis31.

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

1Department of Biotechnology and Bioengineering, Technical University of Denmark, Kongens Lyngby, Denmark. 2US Department of Energy Joint Genome Institute, Walnut Creek, CA, USA. 3US Department of Energy Joint BioEnergy Institute, Emeryville, CA, USA. 4Sandia National Laboratory, Livermore, CA, USA. 5Chemical and Biological Process Development Group, Energy and Environment Directorate, Pacific Northwest National Laboratory, Richland, WA, USA. 6Centre for Structural and Functional Genomics, Concordia University, Montreal, Quebec, Canada. 7Architecture et Fonction des Macromolécules Biologiques, CNRS UMR 7257, Aix-Marseille University, Marseille, France. 8Institut National de la Recherche Agronomique, USC 1408 Architecture et Fonction des Macromolécules Biologiques, Marseille, France. 9Department of Biological Sciences, King Abdulaziz University, Jeddah, Saudi Arabia. 10Fungal Physiology, Westerdijk Fungal Biodiversity Institute and Fungal Molecular Physiology, Utrecht University, Utrecht, The Netherlands. 11Lawrence Berkeley National Laboratory, Berkeley, CA, USA. 12Department of Microbiology, University of Helsinki, Helsinki, Finland. 13Department of Plant and Microbial Biology, University of California, Berkeley, CA, USA. 14Environmental Molecular Sciences Division, Earth and Biological Sciences Directorate, Pacific Northwest National Laboratory, Richland, WA, USA. 15Present address: Amyris, Inc., Emeryville, CA, USA. *e-mail: scott.baker@pnnl.gov; mr@bio.dtu.dk.
additional A. niger isolates (including two previously described as species A. lactofoetatu 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 pipeline15,14 (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 (InterPro35,36). The tree supports previous proposals10,32 that A. lactofoetatus 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 observation35,36 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 genes in section Nigri have already been sequenced, ~30% of the 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 observation35,36 that this species is quite atypical in section Nigri.

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 genomes) (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 relation between the 32 species in section *Nigri*. The black nodes represent homolog families found only in the species branching from the nodes. The white boxes represent the genes unique to the specific species. Stacked histograms of the gene count, the core-genome genes, and unique genes for each species; values shown are total numbers of genes. Dark colors, InterPro annotation; light colors, no annotation.

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 and Supplementary Fig. 1) revealed that only 45% of the genes lack InterPro domains (Supplementary Table 3a), indicating—as would be expected—that the core-genomes 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-genomes (14% compared with 17%, Fig. 2c; Supplementary Tables 3d and 4a). General functions like transporters, regulators, organelle-specific proteins, primary metabolism, and structural proteins 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 of 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, as well as 200 transcription factors (protein kinases) and 60 regulatory proteins (mainly related to functions involved in secondary metabolism, gene regulation, or protein regulation (protein kinases) (Supplementary Table 7). Finding these functions in nearly all sets of section **Nigri** species supports the general validity of the method.

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 nearly all sets of section **Nigri** species (excluding the six *A. niger* isolates, to remove intraspecies redundancy). Surprisingly, we identified only ten domains that were found in all species, which include 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 ones and are therefore critical differentiators 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. tubebingensis* clade (Fig. 2d). The *A. niger* isolates have a high

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**Fig. 3 | Gene family sizes for central metabolism pathways from 32 section **Nigri** genomes and 6 reference species.** Proteins putatively catalyzing the same metabolic step (that is, complex subunits or functionally related enzymes) have been grouped in gray boxes. Species are sorted according to the phylogeny in Fig. 1. **a.** Glycolysis isoenzymes. **b.** Tricarboxylic acid cycle isoenzymes. **c.** Growth of strains on CREA medium, which turns yellow with lowered pH. *A. niger* CBS 513.88 has been mutagenized.
isolates, showing that *A. niger* as the intraspecies variation in the clade is of the same scale interspecies variation in the whole (Supplementary Tables 4c and 8c). In summary, the *Nigri* core-genome (3,798 domains) *A. niger* complex 86%; Supplementary Tables 8a,d and two groups are largely of unknown function (*A. tubingensis*).

The percentage of genes with predicted functions clade (77% shared pan-genome, 7% unique; Supplementary Table 11). However, the growth phenotypes show no correlation with the gene content (Supplementary Fig. 6).

**Carbon utilization is not correlated with carbohydrate-active enzyme content.** Aspergilli have a particularly broad ability to degrade and convert plant biomass. It is thus essential to examine the species diversity of this trait at the genotype and phenotype levels. We predicted the carbohydrate-active enzyme (CAZyme) gene content of the genomes across section *Nigri* (17,903 CAZyme domains; Fig. 4 and 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.

Carbon 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 genomes41. 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, and 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 fungisporein44/nidulanin A45, 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 Fig. 8b). The cluster is found in all A. tubingensis isolates as well as in A. niger (phoenicis ATCC 513.88). 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 Fig. 8b). The cluster is found in all A. tubingensis isolates as well as in A. niger (phoenicis ATCC 513.88). 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 pre-aurantiamin complex and the remain-
species genomes (with the exception of A. tubingensis, A. niger, A. brasiliensis, and A. vadensis) encode one or more unique SMGCs. These patterns show the existence of high diversity of SMGCs between species and of a homogeneous 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 naphth-γ-pyrones, of which aurasperone B\(^2\) 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 graminearum\(^3\). 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 genomes, which adds substantial information on fungal genetic diversity. We further combined genome analysis with metabolite 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.

**URLs**

Carbohydrate-Active enZymes Database, http://www.cazy.org; RoerdamAndersenLab GitHub repositories of scripts for comparative genomics, https://github.com/RoerdamAndersenLab/h; HGAP Assembly, http://files.pacb.com/software/smrtnanalysis/2.2.0/doc/smrtportal/help/1SSLI/Wehtml/CS_Prot_RS_HGAP_Assembly3.htm; JGI fungal genome portal MycoCosm, http://jgi.doe.gov/fungi; JGI website, http://jgi.doe.gov; Joint BioEnergy Institute website, http://www.jbei.org; Technical University of Denmark (DTU) Bioinformatics TargetP web server, http://www.cbs.dtu.dk/services/TargetP/

**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.

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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.H. and M.T.N. engineered strains, analyzed data, and wrote parts of the manuscript. J.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 I.Y.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

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Methods

Fungal strains. Unless otherwise noted, the species examined were taken from the JBT 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) cultivated for 5–10 d at 30 °C. Biomass was isolated by filtering through Miracloth (Millipore, 47.855x1.0), freeze-dried, and stored at 80 °C. DNA isolation was performed using a modified version of the standard phenol extraction (ref. 2 and below) and checked for quality and concentration using agarose gel (BioRad). 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, hypapae 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 Fulton et al. 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 SSII (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 HiSeq 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 to remove artifacts and contamination and assembled using Velvet36. The resulting assemblies were used to create in silico long mate-pair libraries containing poly(T) oligos (PSTC00000000), A. costaricensis (PSTH00000000), A. ellipticus (PSTY00000000), A. eucalypticola (MSFU00000000), A. fijitensis (PSTG00000000), A. heteromorphus (PSTJ00000000), A. homomorphus (PSTO00000000), A. iherbicus (PSTB00000000), A. megasporus (PSTP00000000), A. luteocephalus (MSFR00000000), A. neoger (MSF00000000), A. niger ATCC 13157 (A. phoenicia) (QQUR00000000), A. niger ATCC 13496 (QQJZ00000000), A. piperis (PSTD00000000), A. saccharolyticus (MSQF00000000), A. sclerotiorum (PSSW00000000), A. sclerotiorum (MSKF00000000), A. uvurum (MSFT00000000), A. vadosius (MSFS00000000), and A. weltschii (QQCJ00000000).

See also the Nature Research Reporting Summary linked to this article.

Analysis of secondary metabolism. Cultivation for secondary metabolism analysis. Fungal strains were cultivated as three-point cultures on CYA, CYAS29, 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 methanol37.

Extraction of fungal metabolites. Fungal metabolite extracts were prepared using one of the three following methods38: (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 (UPLC) 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, 2.6 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 database39.

Method 2. UHPLC–DAD–HRMS was conducted on a Dionex RSLC UltiMate system linked to a maxIxs 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, 2.6 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, while 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)29 and some in ESI–full scan mode, scanning m/z 100–1,250. Data were analyzed by aggressive dereplication29 using lists of compounds considered to be from be forkaspergillus 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 Antibase2012 and dereplicated using ascore mass, isotope patterns, adduct patterns, log D, and ultraviolet-visible data39.

Method 3. UHPLC–DAD–HRMS was conducted on an Agilent Infinity 1290 UHPLC system coupled to an Agilent 6550 Q-TOF 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% to 100% acetonitrile over 15 min, where it was held for 2 min. The flow was 0.35 ml min⁻¹ and the temperature 60 °C. Injection volume was between 0.1 and 1 μl, depending on the sample concentration.

Some samples were analyzed in ESI– and some in ESI–full scan mode, scanning m/z 100–1,700 and with automatic MS/MS enabled for ion counts above 100,000 and with a quarantine time of 0.06 min. MS/MS spectra were obtained at 10, 20, and 40 eV (ref. 29). Full-scan data were analyzed as above in MassHunter29. MS/MS data were matched to our internal MS library (~1,700 compounds) of reference standards and tentatively identified compounds29.

Genome annotation and analysis. Genome annotation. All genomes were annotated based on the JGI annotation pipeline34 as previously described60.

Swiss-Prot comparison. Swiss-Prot comparisons were done using protein BLAST alignments with BLASTp (e-value cut-off 1 x 10⁻10). Orthologous groups of sequences were
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 ($-t = p; -b = 45; -i = e = h$). 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 pipeline as described in the studies using InterPro version 69.

According to SMURF, the following genes were predicted as a ‘backbone’ gene set:

- Genes that have at least three PFAM domains—ketocyl-synth (PF00109), Ketoacyl-synth_C (PF02801), and Acyl_transf_1 (PF00698)—were predicted as ‘PKS’ genes.
- Genes that have ketocly-synth (PF00109) and Ketoacyl-synth_C (PF02801) but not Acyl_transf_1 (PF00698) were predicted as ‘PKS-like’ genes.
- Genes that have only the three domains AMP-binding (PF00501), PP-binding (PF00550), and Condensation (PF00668) were predicted as ‘NRPS’ genes.
- Genes that have an AMP-binding (PF00501) domain and at least one of the domains PP-binding (PF00550), Condensation (PF00668), NAD_binding_4 (PF07993), and Epimerase (PF01370) were predicted as ‘NRPS-like’ genes.
- Genes that have both ‘PKS’ and ‘NRPS’ domains were predicted as ‘Hybrid’ genes.
- Genes that have a Trp, DMAT domain were predicted as ‘DMAT’ genes.
- Genes that have Terpene synth (PF03979) or Terpene synth_C (PF03936) domains were predicted as ‘Terpene cyclase/synthase’ genes.

Secondary metabolite-specific PFAM domains were taken from Supplementary Table 2 of the SMURF paper.

As input, the program takes genomic coordinates and the annotated PFAM domains of the predicted genes. Based on the multidomain 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 (NRPs), (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 SignalP4.1.

**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+ suite with e-value cut-off $1 \times 10^{-10}$ 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 other species.

**Detection of encoded CAZymes.** Each Aspergillus protein model was compared using BLASTp to proteins listed in the Carbohydrate-Active enZYmes database (CAZy)64. 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 crysogenum 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 $\leq 1 \times 10^{-10}$). 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 $>80\%$, the size of the alignment between the pairwise sequences was $>50\%$, and the hit must be found in both of the species’ BLAST output (reciprocal hits). Singleton sequences were assigned a family having only one gene member. This allowed for identification of species-unique as well as genes shared by sections, clades, and sub-clades of species. All homologs 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 and 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 $1 \times 10^{-10}$). 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 species in the analysis (ptailoring/pbackbone = sum of percentage BLAST alignment of tailoring/backbone enzymes, respectively; ptailoring/pbackbone = number of tailoring/backbone enzymes with significant hits, respectively; ptailoring/pbackbone = total number of tailoring/backbone enzymes):

$$\text{ptailoring} \times \text{ptailoring} \times \text{pbackbone} \times \text{pbackbone} \times 0.35 + \text{avg}(\text{pbackbone}) \times 0.65$$

To create a cluster similarity score, a combined score of tailoring and backbone enzymes was created. The sum of the BLASTp percent identity (ptailoring/pbackbone) of all hits for tailoring enzymes between two clusters was divided by the maximum amount of the analysis enzyme (ptailoring/pbackbone) and multiplied by 0.35. Then the score for the backbone enzymes was calculated in the same manner but multiplied by 0.65 to give more weight to the backbone enzymes. The scores were added to create an overall cluster similarity score:

$$\text{avg}(\text{pbackbone}) \times 0.35 + \text{avg}(\text{pbackbone}) \times 0.65$$

**Identification of shared SMGC families at nodes of the phylogenetic tree.** A list containing organisms of each branch of the phylogenetic tree was created and compared with the list of organisms for each SMGC family. If all organisms of a family matched, the count on the corresponding node was increased by one.

**Prediction of the aurasperone B gene cluster.** Lists of organisms for all SMGC families were compared with the lists of aurasperone B-producing species and filtered for InterPro annotations containing the terms ‘cytochrome P450’ or ‘methyltransferase’.

**Primary metabolism.** Copy number numbers were assessed using the homologous protein families generated during the analysis of genome diversity. The gene pathway associations were taken from the A. niger genome-scale model41. All proteins in the respective protein families were considered putative isozymes and were included in the copy number analyses.

Comparing the putative isoenzymes in the different species, gene sequences were aligned and clustered using neighbor-joining with MUSCLE v3.8.314. Resulting trees were visualized and edited for publication using the Python ETE Toolkit47. Subcellular localizations for the genes included in the analysis were predicted using the TargetP™ web server (see URLs).

**Reporting Summary.** Further information on research design is available in the Nature Research Reporting Summary linked to this article.

**Code availability.** Code for generation of gene families of homologs and for generation of SMGC families is available from GitHub (see URLs).

**Data availability.** All genomes used in the study are available from Joint Genome Institute fungal genome portal MycoCosm (http://jgi.doe.gov/fung). All new genomes published in the study have been deposited in the DNA Data Bank of Japan (DDBJ)/European Molecular Biology Laboratory (EMBL)/GenBank under the following accession: A. aculeatinus (PST0000000000), A. brunneoviolaceus (PST0000000000), A. casticaricus (PST0000000000), A. ellipticus (PST0000000000), A. eucyrtiaca (PST0000000000), A. fijenesis (PST0000000000), A. heteromorphus (PST0000000000), A. homomorphus (PST0000000000), A. iibericus (PST0000000000), A. indolensis (PST0000000000), A. japonicus (PST0000000000), A. laeticolor (PST0000000000), A. neomiger (PST0000000000), A. niger ATCC 13157 (A. phoenicis) (QQR0000000000), A. niger ATCC 131496 (QQR0000000000), A. piperis (PST0000000000), A. saccharolyticus (PST0000000000), A. sclerotioniger (PST0000000000), A. theobromae (PST0000000000), A. violaceofuscus (PST0000000000), and A. weisitae (QQR0000000000).
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| Sample size | N/A |
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| Data exclusions | N/A |
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