Comparative genomic analyses reveal the features for adaptation to nematodes in fungi

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1. Introduction

The pine wood nematode is responsible for an epidemic of pine wilt disease, which causes severe ecological and economic losses in Asia and Europe and more recently spreading to Mexico.1,2 Esteya vermicola (Ev) is the first recorded endoparasitic nematophagous (NP) fungus that attacks the pine wilt nematode (PWN; Bursaphelenchus xylophilus), which exhibits high infectivity.3 Certain volatile organic compounds (VOCs), α-pinene, β-pinene, and camphor for PWN attraction were produced by Ev.4 Then lunate adhesive spores produced by Ev, whose concave surface has large surface area and thick adhesive layer, adhered to the cuticle of attracted nematode, and penetrated the cuticle of PWN. Afterwards, Ev colonized in the body cavity of the
nematode and produced a great deal of mycelia and lunate spores on cadaver of nematode to attract new nematodes for next infection cycle\(^5\) (Fig. 1). Once lunate spore adheres to nematode cuticle, spores fall-off from the conidiophores and move with nematode. Ninety percent of inoculated PWN are infected within 24 h.\(^3,6\) The high infectivity suggests that \(E.v\) has great potential as biocontrol agent against PWN, but the evolution and mechanisms underlying infection are not clear.

NP fungi are important microorganisms that can suppress the populations of plant-parasitic nematodes and maintain ecological balance and can be used as potential biological control factors.\(^7\) The genomes of NP fungi which have been sequenced include \textit{Arthrobotrys oligospora} (\(Ao\)) with adhesive networks,\(^8\) \textit{Dactylellina haptotyla} (\(Dh\); synonyms \textit{Monacrosporium haptotylum}) of knob-forming species,\(^9\) \textit{Drechmeria coniospora} (\(Dc\)),\(^10,11\) \textit{Pochonia chlamydosporia} (\(Pc\))\(^12,13\) and \textit{Hirsutella minnesotensis} (\(Hm\)) of endoparasitic fungus,\(^14\) and \textit{Drechslerella stenobrocha} (\(Ds\)) with constricting ring.\(^15\) Despite the important roles of NP fungi in ecology, comprehensive and detailed analysis of similarity and difference based on the multiple genomic level was still lacking. Due to very different hosts categories and lifestyles of insect pathogenic (IP), plant pathogenic (PP) and NP fungi, it has been expected that their capacity to utilize the host’s nutrition and to degrade carbohydrates and proteins would also differ, and that might be reflected in their genetic makeup and gene copy numbers at first. The availability of multiple whole genome data of NP, IP, and PP fungi has provided good opportunity to employ genome-wide surveys to investigate how natural selection shaped the evolution of NP fungi towards nematode host.

The study aimed to decipher the putative genetic components involved in pathogenesis and evolution and to identify the molecular underpinnings underlying the establishment of predation relationship in NP fungi by performing comparative genomic and evolutionary analysis. Here, we reported the complete genome sequence of \(E.v\) based on the PacBio sequencing technology, comprehensively analyzed the differences among NP, PP, and IP fungi, and uncovered the genetic and evolutionary bases of NP fungi (\(Ao\), \(Dh\), \(Dc\), \(Hm\), \(Pc\) and \(E.v\)) by the numbers of pathogenic genes, secreted protein, secondary metabolism encoding genes, expanded gene families, and genes undergoing positive selection.

2. Materials and methods

2.1. Fungal strain and DNA preparation

\(E.v\) (CBS115803) was purchased from CBS and maintained on cornmeal agar. This fungus was originally isolated from \textit{Scolytus intrica-tus} and its galleries in oak trees in Czech Republic.\(^16\) Mycelium was grown in liquid medium (potato dextrose broth) and incubated at 25 °C using a shaker at 150 rpm for 7 days. The mycelium (2g) was harvested by aseptic filtering and then grounded in liquid nitrogen. Genomic DNA was extracted using Qiagen Genomic-tip Kit 500 G (Cat No./ID: 10262, Germany) according to the manufacturer’s instructions. DNA quality was assessed by spectrophotometry and gel electrophoresis before library construction.

2.2. SMRT PacBio sequencing and HGAP genome assembly

Twenty-five micrograms of high-molecular-weight \(E.v\) gDNA was used for the PacBio libraries constructing. A 20-kb insert SMRTbell
library was generated using a 15 kb lower-end size selection protocol on the BluePippin (Sage Science). The genome was sequenced using 4 SMRT Cells and P6-C4 chemistry on the PacBio RS II platform (Pacific Biosciences). The *Ev* genome was assembled using the HGAP (version 2.3.0). First, the genome was preassembled with improved consensus accuracy. Then it was assembled through overlap consensus accuracy using Celera assembler and polished with Quiver to improve the site-specific consensus accuracy of the assembly. This Whole Genome project has been deposited at SRA database in NCBI under the accession SRP097009 (BioProject: SUB2318437; BioSample: SUB2318464).

2.3. Gene prediction and functional annotation
Gene annotation was conducted by combining *de novo* and homology-based methods. Augustus and GeneMark-HMM were employed to *de novo* predict gene structures in the *Ev* genome, respectively. For homology-based prediction, protein sequences from three species ([*Grosmania clavigera* (Gc), *Ophiostoma piceae* (Op), *Sporothrix schenckii* (Scb)]) were initially mapped onto the *Ev* genome using BLASTn (E-value ≤1e−5). The homologous genome sequences were aligned against the matching protein using GeneWise for accurate spliced alignments. All the predictions were combined by EVIDenceModeler (EVM) to produce a consensus gene set. Gene functions were assigned according to the best matches derived from the alignments to proteins annotated in KOG, NR, SwissProt, and TrEMBL databases using Blastp (E-value ≤1e−5). Then the pathway in which the gene might be involved was annotated by KAAS according to the KEGG database. Motifs and domains were annotated using InterProScan by searching against GO databases. Finally, the result annotated from the KOG, GO, KEGG, NR, SwissProt, and TrEMBL databases were combined to obtain the final annotation of *Ev* genome.

2.4. Protein family classifications and prediction of pathogenicity-related genes
The carbohydrate active enzymes (CAZy) genes were identified using Blastp according to the CAZy database. Whole genome protein families were classified by Pfam analysis. The families of proteases were identified by Blastp searching against the MEROPS peptidase database release 9.4.28 Transporters were retrieved from the Transporter Classification Database (TCDB). Putative virulence factors were identified by searching against the pathogen-host interaction database (PHI). database of virulence factors in fungal pathogenes (DFVF), database of virulence factors of bacterial pathogens (VFDB). Peroxidase-encoding genes were predicted searching against the MEROPS peptidase database. Antimicrobial resistance genes were identified according to Antibiotic Resistance Genes Database (ARDB). An E-value cut-off of 1.0e-5 was adopted to filter the BLAST results of the above databases. Twenty-one strains of fungi were used for the above BLAST analysis (Supplementary Table S1). SignaP 4.1 and TargetP were used to predict potential secreted proteins with default parameters. Secondary metabolite genes of six NP fungi, six IP, and six PP fungi were predicted by SMURF (Supplementary Table S1). The antiSMASH pipeline with HMM signatures was used to identify and annotate putative polyketide synthase (PKS), nonribosomal peptide synthetase (NRPS), and terpene synthase (TPS) genes and other gene clusters about secondary metabolites and to predict the PKS and NRPS domain architecture. The whole genome dataset was subjected to analysis with default settings. The correlations between these 18 fungal species (IP, PP, and NP) and gene variables (PHI, TCDB, DFVF, and VFDB) were evaluated by correspondence analysis. All statistical analyses were performed with R project V.3.4.2.

2.5. Phylogeny construction and estimation of divergence time
Twenty-three fungi species ([*Ev*, *Dc*, *Ao*, *Db*, *Hm*, *Pc*, *Gc*, *Op*, *Ss*, *Sporothrix brasiliensis* (Sb), *Fusarium graminearum* (Fg), *Magnaporthe oryzae* (Mo), *Colletotrichum higginsianum* (Ch), *Verticillium dahliae* (Vd), *Zymoseptoria tritici* (Zt), *Cordyceps militaris* (Cm), *Beauveria bassiana* (Bb), *Metarhizium anisopliae* (Ma), *Metarhizium robertsi* (Mr), *Moelleriella libera* (ML), *Sporothrix insectorum* (Si), *Neurospora crassa* (Nc), *Saccharomyces cerevisiae* (Sc)] were used for gene family clustering and phylogeny tree construction. Blastp was used to generate the pairwise protein sequence with similarity (E-value ≤1e−5). OrthoMCL was used to cluster similar genes by setting main inflation value 1.5 and other default parameters; 1203 single-copy gene families were extracted for the phylogenomic analysis. The protein sequences were aligned by Mafft for accurate spliced alignments. All the predictions were combined by EVidenceModeler (EVM) to produce a consensus gene set. Gene functions were assigned according to the best matches derived from the alignments to proteins annotated in KOG, NR, SwissProt, and TrEMBL databases using Blastp (E-value ≤1e−5). Then the pathway in which the gene might be involved was annotated by KAAS according to the KEGG database. Motifs and domains were annotated using InterProScan by searching against GO databases. Finally, the result annotated from the KOG, GO, KEGG, NR, SwissProt, and TrEMBL databases were combined to obtain the final annotation of *Ev* genome.

2.6. Expansion and contraction of gene families
Gene families that have undergone expansions or contractions were identified using CAFE version 3.0 program with default parameters. The algorithm CAFE takes a matrix of gene family sizes in extant species as input and uses a probabilistic graphical model to ascertain the rate and direction of changes in gene family size across the given phylogenetic tree. The tree and expansion/contraction data were displayed using the iTOL web tool.

2.7. Identification of orthologous genes and positively selected genes (PSSGs)
Orthologous genes in 23 species were identified. Genes with CDS length less than 450 bp were filtered. All-versus-all Blastp comparisons were used to identify clusters of homologous genes. For avoiding noise information in the Blastp output, orthologous clusters that contained more than 10 species were retained. The values of Ks and Ka substitution rates and the Ka/Ks ratio were estimated using the Codeml program in the PAML package. Codeml was used to calculate the selection pressure in the phylogenetic tree (P < 0.05).

3. Results
3.1. General features of *Ev* genome
*Ev* genome was sequenced at approximately 100 x coverage, generating a total of 4.44 Gbp sequences. *De novo* assembly yielded *Ev* genome of 34.2 Mb in length with average 68.59 x coverage depth.
3.2. Comparative analysis of gene involved in pathogenicity and virulence

To identify functional gene categories characteristic of adaptation in killing nematode, we compared 21 genomes, including 6 IP fungi, 6 PP fungi, and 6 NP fungi and 3 genetically close species with Ev, Gc, and Si (Supplementary Table S1). The genome integrity and completeness of gene prediction assessed by BUSCO are shown in Supplementary Table S2. The genome integrities of the selected species were more than 95%, and the completeness of protein prediction was greater than 92%. The corresponding datasets based on different databases were shown in supporting information (Supplementary Table S3–S11). Extracellular adhesive proteins comprise GLEYA proteins, CFEM proteins, WSC-containing proteins and lectin, each of which was averagely the highest in NP genomes except CFEM (Supplementary Fig. S1), as well as total adhesion (Supplementary Fig. S2). Dh encoded the highest number of adhesive proteins GLEYA and WSC, whereas Ao encoded most abundant lectin among 21 analyzed fungi genomes (Supplementary Fig. S1a). However, our comparative analysis found that endoparasitic fungi (Ev, Dc, Pc and Hm) contain fewer extracellular adhesive proteins (Lectin and GLEYA) than adhesive networks producer Dh more subtilisin S8 and subtilisin-like were found in Agely abundant in NP (Supplementary Fig. S3 and Supplementary Table S6). More subtilisin S8 and subtilisin-like were found in Ao and Db than that in other NP fungi (Supplementary Table S6). NP fungi encoded averagely richer ankycin repeat protein, while PP fungi encoded more leucine-rich repeat, tyrosinase, xylanase (Supplementary Fig. S4). IP fungi encoded the least cellulose, compared with PP, NP and CP fungi (Supplementary Fig. S6). Interestingly, all analyzed fungi except Z. tritici possessed appressorial penetration protein (PHE: 256), homologs of proteins in plant-pathogenic fungi Mo, especially in Ao (25) and Db (28) most abundant (Supplementary Table S4).

Chitinase is important for IP and egg-parasitic fungi to infect insect and nematode eggshell.45-46 IP fungi were most abundant averagely in enzymes catalyzing the decomposition of chitin (Supplementary Fig. S7). Px, an egg-parasitic NP fungus, encoded the highest number of chitinase (159, CAzy) among NP fungi (Supplementary Fig. S7). Chitosanases (including GH5, GH7, GH8, GH46, GH75 and GH80) are another enzymes involved in chitin degradation and nematode parasitism.13 The analysis suggested that Ev had the most chitosanases, especially GH7; Ao encoded the richest GH75 (Supplementary Fig. S3 and Supplementary Fig. S7). IP, NP, and PP fungi also contained averagely more peptidase than CP fungi (Supplementary Fig. S8). IP fungi possessed most pkinase and enterotoxin-a, furthermore, and both pkinase and enterotoxin-a in IP fungi was significantly higher than that in PP fungi (Supplementary Figs S9 and S11). Based on Pfam database, IP had significantly more toxin genes than PP and CP fungi (Supplementary Fig. S10). The number of aspartic peptidase in NP and IP fungi was averagely higher than PP fungi (Supplementary Fig. S5). PP fungi genome encoded the richest CAZy enzymes (Supplementary Table S3), peroxidase (Supplementary Table S9), in particular, hemicellulose (Supplementary Fig. S12), G-protein receptor (Supplementary Fig. S12), pectate lyase (Supplementary Fig. S13) and cutinase (Supplementary Fig. S14), P450 (Supplementary Fig. S15). The important claim here was that the number of hemicellulose, G-protein receptor, and pectate lyase in PP fungi was significantly higher than that in IP fungi (Supplementary Figs S12b and S13b). Moreover, the content of cutinase in PP group was significantly greater than that in IP and CP group (Supplementary Fig. S14b). It suggested that PP fungi evolved to adapt to higher carbohydrate content in plants.

We identified 231 transporter superfamily in all 21 fungi species. Ev, Hm and Pc genome encodes a large number of transporters (1,482, 1,546, and 1,986, respectively; Supplementary Table S10). Remarkably, Ev and Pc had more carbon (72, 90), iron (23, 19) and vitamin transporters (81, 53) than other NP fungi (Fig. 2 and Supplementary Table S10). The numbers of genes encoding allantioate transporters, and vitamin or iron in Ev were the greatest and second highest among all 21 fungi, respectively (Fig. 2). Ammonium and urea import capacity of Ev was higher than other NP, IP, and PP fungi. Therefore, Ev shows an increased genetic potential in terms of vitamin, iron, and nitrogen uptake. This could be an important mechanism for perturbing nematode metabolism and ultimately causing nematode death. It indicated that the transport capacity of ammonium, nucleo-base, and nucleoside in CP group was very strong (Supplementary Fig. S16). The number of oligopeptide transporter in IP group was averagely higher than other groups and significantly higher than that of in NP group (Supplementary Fig. S16). This implied the great potential of absorbing oligopeptide of nitrogen sources in IP fungi.

Strikingly, all 21 fungi contained a wide array of antibiotic resistance genes. The total number of antibiotic resistance genes in IP fungi (207) was the richest, followed by PP fungi (201), and NP fungi (174) (Supplementary Table S8 and Supplementary Fig. S17). The substances encoded by these genes were likely to play roles in competition with bacteria in insect, nematode, and plant, then pathogenic fungi can colonize in their host. Among antibiotic synthesizing genes, macrolide transporter-encoding genes and multidrug-encoding genes were most abundant in each group (Supplementary Table S8 and Supplementary Fig. S17).

3.3. Secreted proteins and small secreted proteins

Secretion signals were predicted in 21 fungi. The results showed that PP had the highest number of secreted proteins (SPs) and small secreted proteins (SSPs; < 300 amino acids), followed by NP and IP fungi (Supplementary Table S12). Among the 21 analyzed fungi, Mo had amazing ability to produce SSP (about one-tenth of the total protein). Among NP fungi, Pc had the most SPs and SSPs, followed by Dh, Ao and Hm (Supplementary Table S12). Out of the 1,318 proteins predicted to be secreted by Ev, 46.7% can be ascribed known function. Of these 397 SSPs of Ev, 302 was annotated into NR database, but only 131 (33.0%) were annotated into known function; 10 of 397 SSPs was related with transporters; 48 of 397 SSPs (12.1%) was related with pathogen-host interaction based on Pfam database; and 180 of 397 SSPs (45.3%) was found listed in Pfam database.

The possible pathogenic-related SPs are shown in Fig. 3. NP fungi were rich in SPs of subtilisin or subtilisin-like protease, lectin, WSC, and iron transporter. Only Ev was predicted to produce secretory celllobiose dehydrogenase in NP fungi. Ev was remarkable in SPs of putative aspartic protease, glucanase and endoglucanase, laccase and celllobiose dehydrogenase. It was worth noting that the number of secreted oligopeptide transporter and aspartic proteinase was remarkably high in endoparasitic fungi Ev, Dc, Pc, and Hm. The important

The N50 sizes of 50 contigs and 42 scaffolds were 4,26 Mbp and 4,41 Mbp, respectively. Combined de novo and homology prediction produced 8,427 protein-coding genes. The detailed annotation results are given in the supporting information.
SPs involving pathogenicity in Ev were comparable to those in Mo and Ch, such as P450, laccase, and cellobiose dehydrogenase.

3.4. Secondary metabolite analysis
PP fungi can produce diverse secondary metabolites (SMs) that aid in pathogenicity. However, the extent and distribution of the capacity for secondary metabolite synthesis have not been investigated in NP group. Furthermore, there was no comparative analysis about the capacity for secondary metabolite synthesis among groups (IP, NP, and PP) before. SMURF predicted that endoparasitic fungi (Hm, Pc, Dc, and Ev) had more genes encoding secondary metabolites than species of adhesive networks and knob-forming species (Ao and Dh) (Supplementary Table S13). Hm had the highest number of gene clusters and core genes involved in biosynthesis of secondary metabolites, followed by Pc among NP group. PKS and NRPS synthesis genes in Pc (195, 156) and Hm (299, 238) were more abundant than that of other analyzed NP fungi. Especially endoparasitic NP fungi had tremendous potential for biosynthesis of secondary metabolites that act as toxins or signals in the interactions between fungus and host. The rich unknown secondary metabolism in NP fungi is worthy of further research and exploration. Furthermore, Mr and Ma of IP fungi, and Ch and Mo of PP fungi also had great capacity for biosynthesis of secondary metabolites, which possibly greatly enhanced their disease-causing abilities.

3.5. Insights into pathogenic differentiation leading to different hosts categories
To assess pathogenic differentiation leading to different host classification, we performed correspondence analysis based on multiple large datasets (Supplementary Tables S3–S11). It was hypothesized...
that pathogenic fungi belonging to the same host categories with similar nutritional mode, would be clustered together. Figure 4 shows separation result of three clusters (PP, IP, and NP) according to gene variables of PHI genes, membrane transport proteins (TCDB), fungal virulence factors (DFVF), virulence factors of bacterial pathogens (VFDB). The species of IP fungi were clustered except Si species, and so did the six PP species. However, the distribution of NP species was dispersive, and Ev, Dc, and Pc were always clustered with PP, IP fungi, respectively. CAZy, peroxidises, and antibiotic resistance genes failed to separate IP and PP fungi (figures not shown). Also, the analysis suggested that the difference in pathogenicity genes or membrane transport proteins was possibly the important factors to drive pathogenic differentiation and adaptation to different host categories. Endoparasitic fungal Ev, Dc, Pc, and Hm, which were very different from Ao and Dh, were likely to be the intermediate type between the IP and PP fungi.

3.6. Phylogeny and divergence time estimates
OrthoMCL was used to identify gene orthologs. Totally 253,363 genes from 23 species were used for gene family clustering analysis. Finally, 20,524 gene families containing 206,300 total genes from 23 species were generated, out of which 1,203 possessed a strict single-copy orthologous relationship. The phylogenomic tree was constructed using concatenated nucleotide sequences of 1,203 single-copy orthologous genes to infer NP fungal phylogenetic relationships in the context of other fungal taxa. It indicated that Ev diverged from plant-associated Gc about 135 million years ago (Mya), and Ev and Gc had closer genetic relationship (Fig. 5). The phylogenetic analysis showed that the relationship of Ao and Dh in NP fungi were closely related, and the divergence of the two species took place around 212 Mya. Dc, Pc and Hm had a close genetic relationship with IP fungal Metarhizium, which was consistent with previous phylogenetic analysis results.10–14

3.7. Gene family and gene family expansion
Gene duplication and protein family expansion are important genomic mechanisms that shape the evolution of pathogenic fungi.59 CAFE analysis was performed to assess the gene family variations; 276 gene families were expanded in Ev, 1,170 families had undergone contraction (Fig. 6). We found a marked expansion of transporter genes in Ev. Among the 276 expanded gene families in Ev, 10.7% (37) was related to transporters or permeases, which was the highest among all the fungi analyzed (Supplementary Table S14), while the proportions were less than 2.4% and 6.8% in other five NP fungi and other analyzed species, respectively (Supplementary Table S14). It was surprising that the proportion of expanded gene families of transporters in Ev was much higher than that of the other analyzed species. The transporters of expanded gene families in Ev occurred mainly for the carbon source, nitrogen source, iron and vitamin transport, such as ABC transporter, sugar transporter, iron permease, urea transporter, allantoate/allantoale permease, vitamin H transporter and amino acid permease (Supplementary Table S15), which provide the basis for effectively absorbing nutrition of nematodes. Gene families involving in aflatoxin efflux pump had undergone expansion in Ev (Supplementary Table S16). Gene families about endo-beta-glucanase, which catalyse hydrolyzing O-glycosyl compounds, were substantially expanded in Ev compared with other fungi species. Important pathogenicity-related gene families were

Figure 4. Correspondence analysis grouping pathogenic fungi by host class based on the set of predicted (a) PHI, (b) DFVF, (c) TCDB, (d) VFDB.
Figure 5. Phylogenetic tree and divergence time estimates. The numbers at nodes in the tree are the divergence times from the present (Million years ago, Mya). The nodes with red dot represent the reliable calibration time of the divergence. Bootstrap values in each node are 100 (not shown in the diagram). Six NP fungi H. minnesotensis, D. coniospora, P. chlamydosporia, E. vermicoila, A. oligospora, and D. haptotyla.

Figure 6. Gene family expansions and contractions. Branch length represents differentiation time. Numbers for expanded (green) and contracted (red) gene families are shown below branches or taxon names with percentages indicated by pie charts. Blue represents the proportion of gene families no changed in the current species or branch. Orange in pie charts represents the proportion of the gene families combined change across lineages. Six NP fungi are marked with red.
founded expanded, including peptidase S41, cellobiohydrolase and laccase.

The six NP fungi in total shared 3,717 gene families, but no expansion gene families in common were observed (Fig. 7a and b). Surprisingly, we found that Ao and Dh shared the largest set of gene families (8,078) through pair comparison, followed by Pc and Hm with 6,807 gene families. Beyond that, the two species (Pc and Hm) with similar nematocidal mechanism (endoparasitism) shared the most expanded gene families (132). Substantial differences were observed in species-specific gene families among six fungi (Ev 1,027, Hm 650, Ao 244, Dh 223, Dc 168, Pc 1,581) in magnitude. The ratio of species-specific gene family to the total gene family was the highest in Pc (16.6%), followed by Ev (14.2%) and Hm 8.1% (Supplementary Table S17). The ratio of species-specific expanded gene family to the total expanded gene family was 71–86.8% among NP fungi (Supplementary Table S17). NP fungi contained more gene families not found in other pathogenic fungi (Supplementary Fig. S18) indicating their unknown functional diversification.

We further compared expanded gene families in 23 species to detect gene family evolution (Supplementary Table S16). Subtilisin-like gene family in NP fungi (Dh and Hm) and IP fungi (Mr) were expanded, except that in PP species. Genes encoding chinatinase, aspartic proteinase and Wsc domain protein had undergone expansion in NP and PP group, but not in IP group. Furthermore, peptidase S41 family protein exhibited expansion in NP fungi (Ev and Dc) and IP fungi (Ma). Intriguingly, gene families about trehalose metabolism expanded greatly in Hm. Sabine Fillinger et al. found trehalose rapidly degraded upon induction of conidial germination and was required for the acquisition of tolerance to a variety of stresses in Aspergillus nidulans.51 Compared with other analyzed species, there was a major expansion in gene families related to allantoate and iron transport, laccase and endoglucanase in Ev (Supplementary Table S16). The ratio of expansion gene families to total gene families in Hm and Pc peaked at an incredible 11.0% and 8.8%, respectively, while that in other four NP fungi was no more than 4% (Supplementary Table S17), which partly explained the differences in genomic size of these species.

3.8. Positive selection

Positive selection, acting in the evolution of functionally important gene families, is an important driving force for pathogens in adaptation to a pathogenic lifestyle and host specialization.52–54 Identifying targets of positive selection will help to annotate the genome functionally and to elucidate evolutionary processes. Signatures of positive selection specify functionally important regions of the genome.55 Based on the established phylogeny and 10,610 high-confidence orthologues identified among 23 species, positively selected genes (PSGs) of six NP fungi and PP fungal Gc was detected. Dh and Ao had more PSGs (256, 247) than endoparasitic NP (Ev 234, Dc 102, Hm 80, Pc 64) (Supplementary Table S18); 97 PSGs in Ev, this was the highest count among 7 analyzed species, belonged to putative

Figure 7. Venn diagrams and selected orthologous genes undergoing positive selection in six NP fungi. (a) Venn diagram of gene families; (b) Venn diagram of expanded gene families; (c) Venn diagram of orthologous genes of positive selection; (d) Selected orthologous genes of positive selection. ‘1’ means undergoing positive selection; ‘0’ means no positive selection.
4. Discussion

4.1. NP fungi encoded rich secondary metabolites and antibiotics

Secondary metabolites involving in host–pathogen interactions may protect the fungus from predation and allow an organism to survive in its ecological niche. Secondary metabolisms of beauvericin, and bassianolide, and destruxin in insect pathogens were insecticidal. Similarly, secondary metabolites in NP fungi can kill nematodes. Endoparasitic fungi had very rich repertoires of secondary metabolism genes compared with nematode trapping fungi implying endoparasitic fungal great potential for killing nematodes. One of the most important questions about the ecology of NP fungi is how they defend the nematode cadaver against different microbial competitors. NP fungi, especially Pc and Hm, encoded a large number of polyketide and nonribosomal peptide synthases which could be part of the biosynthetic pathway of antibiotics. Besides, the phenomenon that NP fungi possess a set of antibiotic resistance genes is a general mechanism responsible for resistance to antibiotics from other microbes. It was suggested that antibiotic and fungal secondary metabolites displayed antagonistic effect against other microbial competitors during colonizing nematode cadaver.

4.2. The abundance and evolution of pathogenic genes in NP fungi

Adhesive proteins accumulating on the outside surface of adhesive traps or spores of NP fungi play vital roles for those fungi to adhere to nematode cuticles. Adhesive materials may harbour many se- 
traps or spores of NP fungi play vital roles for those fungi to adhere to nematode cuticles. Adhesive proteins accumulating on the outside surface of adhesive
proteases as SPs underwent expansion and positive selection in NP group. Aspartic protease, which was expressed by many pathogenic 
ascomycetes during infection, was subjected to positive selection during the evolution in NP fungi. Nitrogen permease regulator, as well as some proteolytic enzymes, was under forward selection, implying the importance of nitrogen sources for NP fungi. Cytochrome P450s play various roles in secondary metabolism and involve in biodegradation of lignin and various xenobiotic compounds. Our results indicated that PP and IP contained richer P450s than NP group. Nevertheless, P450 and P450 reductase underwent positive selection in NP fungi but not in PP fungal Gc. Therefore, P450s should be functionally important for NP in adaptation to pathogenic lifestyle.

4.3. Abundance of transporter in Ev genome

We identified abundant transporters, which underwent expansion in Ev. Monosaccharide transporter (Hxt1), also as sensor, is required for virulence of the maize pathogen Ustilago maydis and is important for fungal development during the pathogenic stage of the fungus. Pyruvate uptake transporter (BcMctA) is also required for pathogenicity in Botrytis cinerea, and disruption of BcMctA significantly reduced the virulence of B. cinerea on cucumber and tomato leaves. Phytopathogenic bacteria and fungi can use iron uptake systems to multiply in their hosts and to promote infection, and iron transporters have been found to play a critical role in virulence of both animal- and plant-pathogenic fungi, such as siderophores of Aspergillus fumigatus and high affinity iron permease of Rhizopus oryzae, which play critical roles in the expression of virulence. Similarly, these transporters could well take part in pathogenicity during infecting nematodes. In addition, based on a wealth of transporters (Supplementary Table S10), it was possible that NP fungi, especially endoparasitic fungal Ev, Pc, and Hm could be more efficiently utilize nematode nutrients compared with other NP fungi.

4.4. Endo-beta-glucanase in Ev genome was highly expanded

Comparative analysis revealed endo-beta-glucanase encoding genes in Ev genome were highly expanded. Endo-beta-glucanase likely plays a role in degrading glucans polymers from themselves cell wall or their host nematodes. Endo-beta-glucanases were identified as parasitism genes in plant-parasitic nematode. Extracellular glucanase secreted by Acremonium perisicum was considered to participate in the degradation of an extracellular storage glucan. Glucanases produced by some Trichoderma species has been recognized as the key enzymes in the lysis of cell walls during their myco- parasitic action against phytopathogenic fungi. Highly expanded endo-beta-glucanase in Ev genome possibly functions as parasitism, but further research is needed to determine the possibility.

4.5. No α- and β-pinene monoterpenoid synthase-encoding genes in Ev genome

Interestingly, the host pine of PWN, vector Monochamus alternatus of PWN, and NP fungal Ev, belonging to three different kingdoms, all attract nematodes through volatile chemicals: α- and β-pinene, due to Ev’s capacity to attract PWN by volatile chemicals α- and β-pinene and the fact that Ev host intracellular bacteria, it remains vague whether the fungus itself produced the compounds. Annotation results of Ev genome revealed that there were no α- and β-pinene monoterpenoid synthase-encoding genes.
Here we hypothesize that the endobacteria produce the monoterpenoid that attracts PWN. The next step for whole genome sequencing of this endobacteria will further verify the authenticity of the endobacteria producing the monoterpenoid that attracts nematodes.

5. Conclusions
We sequenced the genome of *Ev* and embarked upon a comparative genomics analysis. These results indicated that NP fungi was distinguished from other pathogenic fungi in terms of pathogenic genes sets. Despite genetic predispositions for killing nematode, the microscopic mechanism was different in NP fungi. Our analysis elucidated a repertoire of genes in endoparasitic NP fungi distinct from those of other NP fungi. The availability of the first *Ev* genome sequencing, annotation and evolution analysis was an important advance in helping to decipher its pathogenesis. The results would facilitate our comprehension of the genomic features of NP fungi for adaptation to nematodes. Furthermore, the results also set out a solid theoretical foundation for further research and application of these NP fungi as biological control agents.

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Conflict of interest
The authors declare no conflict of interest.

Supplementary data
Supplementary data are available at DNARES online.

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