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Sellers, Graham S., Jeffares, Daniel C. [orcid.org/0000-0001-7320-0706](https://orcid.org/0000-0001-7320-0706), Lawson, Bex et al. (2 more authors) (2021) Identification of individual root-knot nematodes using low coverage long-read sequencing. PLOS one. e0253248. ISSN: 1932-6203

<https://doi.org/10.1371/journal.pone.0253248>

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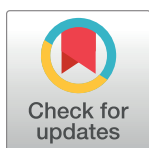
## RESEARCH ARTICLE

# Identification of individual root-knot nematodes using low coverage long-read sequencing

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## Abstract

Root-knot nematodes (RKN; genus *Meloidogyne*) are polyphagous plant pathogens of great economic importance to agriculturalists globally. These species are small, diverse, and can be challenging for accurate taxonomic identification. Many of the most important crop pests confound analysis with simple genetic marker loci as they are polyploids of likely hybrid origin. Here we take a low-coverage, long-read genome sequencing approach to characterisation of individual root-knot nematodes. We demonstrate library preparation for Oxford Nanopore Technologies Flongle sequencing of low input DNA from individual juveniles and immature females, multiplexing up to twelve samples per flow cell. Taxonomic identification with Kraken 2 (a *k-mer*-based taxonomic assignment tool) is shown to reliably identify individual nematodes to species level, even within the very closely related *Meloidogyne incognita* group. Our approach forms a robust, low-cost, and scalable method for accurate RKN species diagnostics.

## OPEN ACCESS

**Citation:** Sellers GS, Jeffares DC, Lawson B, Prior T, Lunt DH (2021) Identification of individual root-knot nematodes using low coverage long-read sequencing. PLoS ONE 16(12): e0253248. <https://doi.org/10.1371/journal.pone.0253248>

**Editor:** Tushar Kanti Dutta, Indian Agricultural Research Institute, INDIA

**Received:** June 7, 2021

**Accepted:** October 27, 2021

**Published:** December 1, 2021

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**Data Availability Statement:** All sequencing data have been submitted to the International nucleotide sequence databases with BioProject accession number PRJNA706653.

**Funding:** This project was internally funded by the University of Hull's Energy and Environment Institute as part of a strategic THYME project.

**Competing interests:** The authors have declared that no competing interests exist.

## Introduction

Root-knot nematodes (RKN; genus *Meloidogyne*) are economically important agricultural pests, parasitising all human crops, and causing major losses to world agriculture [1]. Despite the enormous host range of RKN, perhaps encompassing the diversity of angiosperms [2, 3], not all crops are parasitised by all species. Understanding the species identity of individual infections is important in agriculture since it allows understanding of the threat to different crops, better understanding of natural crop resistance, and design and testing of interventions.

Individual RKN juveniles are part of the soil meiofauna and females, attached to or embedded within the root, corms or tubers, are very small (~350 µm), and none of the life stages are straightforward to identify. Species identification may be carried out using microscopy, or with molecular approaches such as isozyme electrophoresis or PCR diagnostics [4]. It can be time consuming to identify RKN, and these diagnostics often do not scale very advantageously in terms of either time or reagent cost. Lastly, although single sample identification is well established, routine testing and understanding of whole nematode communities in the soil is

much more challenging. The difficulty is not just because of the greatly increased number of sample assignments required, but also because the diversity of species encountered will increase the taxonomic expertise needed, while simultaneously proving difficult for any DNA barcoding or isozyme approach. Extensive surveys of agricultural environments, their existing nematode communities, and the threat to different planting regimes is therefore largely still an aspiration for pest management.

There are over 90 species of RKN [5] and while there is considerable genetic diversity between clades within the genus *Meloidogyne*, there are also species clusters such as the *Meloidogyne incognita* group (MIG) comprising many closely related species including many of the most significant global agricultural pests [6].

The MIG, and a number of other species of RKN, are polyploid apomictic asexual parthenogens. Although cytology suggests a diversity of ploidy levels may exist [7] all sequenced species contain divergent A and B subgenomes as part of this ploidy. These subgenomes may have originated either from interspecific hybridization or else be frozen heterozygosity associated with the loss of sexual reproduction [8–11]. Thus, each species contains within every nucleus divergent genome copies that are more distinct from each other than they are from the same subgenome in a different species. This makes it very challenging to use a single loci for species diagnostics since the A subgenome of *M. incognita* is more similar to the A subgenome of *M. javanica* than it is to the B subgenome in the same *M. incognita* cell [11]. Sampling the orthologous loci is therefore essential to reliable identification and understanding of the species, yet this can be very challenging using single locus PCR assays.

Genomic sequencing approaches are able to capture the evolutionary history of species far better than single locus analyses. Genome assembly using short read sequencing technology is very challenging within the MIG, due to our uncertainties about their ploidy, variations in ploidy between strains and the high levels of intragenomic divergence between genome copies. This can pose significant challenges for assemblers, at present. However, long-read high quality data is producing increasingly complete MIG genomes [12, 13]. Low-coverage long-read Oxford Nanopore Technology (ONT) sequencing can offer a different approach to MIG genomics. For diagnostic analysis, rather than prioritising genome assembly the user could target taxonomic identification, the discovery of novel genetic variability, or the characterisation of virulence-related loci.

Here we develop a method for taxon assignment of RKN using low-coverage, long-read nanopore genome sequencing from single root-knot nematodes. This protocol extracts high quality DNA from juvenile stage 2 (J2), juvenile stage 4 (j4) or immature female individuals, prepares a sequencing library for the Oxford Nanopore Technologies (ONT) Flongle, and allows for multiplexing upto 12 samples per flow cell. From generated sequencing data we use the rapid *k-mer*-based taxonomic assignment tool, Kraken 2 [14]. We show that this cost effective, multiplexed genomic approach can reliably resolve RKN species including the closely related taxa within the MIG.

## Materials and methods

### Nematode culture

Nematode cultures, established from single egg masses, were grown under quarantine licence on tomato plants (*Lycopersicon esculentum*, variety: 'MoneyMaker') for around 3 months at Fera Science, UK. Single females were excised from the root (with sterile hypodermic needles), rinsed with molecular grade water and frozen in sterile 1.5 ml tubes at -20°C before moving to the University of Hull. J2s were hatched by incubating an isolated egg mass at room

temperature in distilled water. Individual j2s were isolated with a dissecting tool, rinsed with molecular grade water and frozen in sterile 1.5 ml tubes at -20°C before moving to the University of Hull.

### DNA extraction

DNA was extracted from individual juvenile stage 2 (j2) or juvenile stage 4 (j4) and immature females excised from root tissue. We employed a modified Mu-DNA: Tissue protocol [15] using a solid phase reversible immobilization (SPRI) magnetic bead capture method [16] to isolate high molecular weight DNA (see [S1 Appendix](#): Section 2 for detailed protocol). This modified Mu-DNA methodology was successful in producing appropriate quality nematode DNA for genomic data. The Mu-DNA extraction technique proved robust and rapid although a standard phenol:chloroform DNA preparation also generated appropriate amounts of DNA in our hands (see [S1 Appendix](#): Section 4).

### Nanopore library preparation and sequencing

Library preparation followed a modified version of the ONT Rapid PCR Barcoding Kit (SQK-RPB004). The kit, although at present designed for MinION, was easily converted for use on the Flongle. DNA yield from a single nematode was invariably much lower than the suggested amount for the protocol. However, increasing both the volume of DNA digested and PCR barcoding cycles produced sufficient template for sequencing from all samples (see [S1 Appendix](#): Section 3 for detailed protocol).

### Basecalling

Guppy basecaller (version 4.2.2) high accuracy (HAC) GPU was used for basecalling and demultiplexing of MinKNOW (version 20.06.05) fast5 output. Basecalling was performed on the University of Hull's Viper HPC on a GPU node (2 x 14-core Broadwell E5-2680v4 processors (2.4–3.3 GHz), 128 GB DDR4 RAM, 4 x Nvidia Tesla K40m GPU). Reads with a minimum average base quality of 7 and with barcodes at both ends of the read were used for downstream analysis.

### Database construction and cleaning

We used a *Meloidogyne*-specific Kraken2 database as all samples in this study had been taxonomically identified as root-knot nematodes prior to sequencing. This database was considerably smaller (6.5 GB) than a database of all nematode genomes (35.4 GB), and as Kraken 2 loads the database to memory this allowed the analysis to be run locally on a modest computer rather than requiring access to a high memory HPC.

Our *Meloidogyne*-specific Kraken 2 database was constructed from all currently available genomic resources for *Meloidogyne* species from Genbank ([Table 1](#)). Despite authors' efforts to remove contaminant sequences during assembly, some may still remain in the final published genomes. To reduce the effect of these potential contaminants in our analysis, we performed a database cleaning procedure. Genome sequences first underwent a Kraken 2 analysis against a bacterial database (constructed from the Kraken 2 bacteria reference library) with a confidence of 0.1. All unclassified reads from this analysis step were subsequently processed against a custom Kraken 2 database (with a confidence of 0.05) of 'contaminants' comprising human (*Homo sapiens*) and a selection of known plant hosts of RKN species; tomato (*Solanum lycopersicum*) and sweet potato (*Ipomoea batatas*). This removed 380 sequences primarily assigned to bacteria, human and plant, resulting in a maximum loss of ~1% of sequences from

**Table 1. Species within the Kraken 2 database.** Eleven species of RKN with 21 total genomes are shown in addition to three non-nematode species included to identify contamination.

Species	Isolate	Accession	Publication
<i>Meloidogyne arenaria</i>	HarA	GCA_003693565.1	[11]
<i>Meloidogyne arenaria</i>	A2-O Okinawa_01	GCA_003133805.1	[17]
<i>Meloidogyne arenaria</i>	Guadeloupe	GCA_900003985.1	[10]
<i>Meloidogyne chitwoodi</i>	Race2	GCA_015183065.1	unpublished
<i>Meloidogyne chitwoodi</i>	Race1	GCA_015183035.1	unpublished
<i>Meloidogyne chitwoodi</i>	Roza	GCA_015183025.1	unpublished
<i>Meloidogyne enterolobii</i>	Swiss	GCA_903994135.1	[12]
<i>Meloidogyne enterolobii</i>	Sample3	GCA_903797545.1	[12]
<i>Meloidogyne enterolobii</i>	L30	GCA_003693675.1	[11]
<i>Meloidogyne floridensis</i>	SJF1	GCA_003693605.1	[11]
<i>Meloidogyne floridensis</i>	JB1	GCA_000751915.1	[18]
<i>Meloidogyne graminicola</i>	VN-18	GCA_014773135.1	[19]
<i>Meloidogyne graminicola</i>	IARI	GCA_002778205.1	[20]
<i>Meloidogyne hapla</i>	VW9	GCA_000172435.1	[21]
<i>Meloidogyne incognita</i>	Kmmt_Gs004	GCA_014132215.1	[22]
<i>Meloidogyne incognita</i>	W1	GCA_003693645.1	[11]
<i>Meloidogyne incognita</i>	Morelos	GCA_900182535.1	[10]
<i>Meloidogyne incognita</i>	Morelos	GCA_000180415.1	[23]
<i>Meloidogyne javanica</i>	VW4	GCA_003693625.1	[11]
<i>Meloidogyne javanica</i>	Avignon	GCA_900003945.1	[10]
<i>Meloidogyne luci</i>	SS-V13	GCA_902706615.1	[13]
<i>Homo sapiens</i>		GCA_000001405.28	
<i>Tomato</i>		GCA_012431665.1	
<i>Sweet potato</i>		GCA_002525835.2	

<https://doi.org/10.1371/journal.pone.0253248.t001>

any genome (S1 Appendix: Section 6). Remaining genome sequences were considered to be free of contaminants and were used for our final database construction.

In addition to the cleaned *Meloidogyne* genomes, we added in a single representative of human, tomato and sweet potato genomes (Table 1). These extra ‘contaminant’ genomes would allow for evaluation of sample and library preparation from plant host and lab based human contamination. The database is provided in the manuscript archive but we do not hold it under git version control due to size.

## Analysis workflow

We built a reproducible data analysis workflow using the Snakemake workflow manager [24]. This workflow uses a conda virtual environment to ensure software compatibility. The basic analysis implements NanoFilt [25] quality control, Kraken 2 [14] taxonomic classification, and presents taxonomic information in textual Kraken 2 outputs and interactive html reports via Recentrifuge [26]. The workflow is available at [https://github.com/Graham-Sellers/RKN\\_genomic\\_taxonomic\\_assignment](https://github.com/Graham-Sellers/RKN_genomic_taxonomic_assignment).

## Taxonomic assignment method development

Taxonomic assignment methods were developed using a test library (RKN\_lib2). This library consisted of 12 *M. incognita* j4 or immature females isolated from root tissue. The samples used ranged from unquantifiable levels of DNA to much higher yields (S1 Appendix: Section

5). We considered this to be a relatively diverse library of sample qualities for method development.

**Quality control and taxonomic assignment.** Sequencing read quality control was performed using NanoFilt, primarily for head and tail cropping of 50 bp to remove library preparation ligation sites. Additionally, NanoFilt assessed the quality of sequencing reads based on a minimum average read quality score. As our Guppy basecaller parameters were for a minimum average read quality of 7, only reads with quality scores of 7 or above were considered for investigation. Kraken 2 has two parameters that can improve accuracy of read classification: confidence (minimum proportion of *k-mers* required for assignment to a taxonomic level) and minimum base quality score (*k-mers* in the read with bases below this value are considered ambiguous and disregarded for assignment). Optimal values for quality control and Kraken 2 classification parameters were determined by their application across samples in the *M. incognita* test library. These optimised values were used for following investigations.

**Read count, length and taxonomic assignment.** We investigated the effect of read number and length on taxonomic assignment and its effect on the proportion of those reads correctly assigned. Reads were pooled from the 12 *M. incognita* samples of RKN\_lib2. For read count, 25 replicates of read count values, at any length, were randomly sampled without replacement from the read pool. For read length, 25 replicates of a set read count for each read length were randomly sampled without replacement from the read pool.

**Display of taxonomy.** Taxonomic assignments are displayed as simplified, *Meloidogyne*-focussed versions of the Recentrifuge html outputs. These were reproduced in R version 4.0.2 [27] using the circlize package [28]. Taxonomy follows the NCBI scheme for which some *Meloidogyne* species have a subgeneric taxonomic group “MIG” while others have only genus and species. This NCBI taxonomy is not completely representative of current *Meloidogyne* phylogenetic relationships [6], but is still fit for purpose for our species diagnostic method. In order to represent this NCBI taxonomy, our figures, and Recentrifuge when focused on *Meloidogyne*, display three concentric rings: the inner one is genus, the outer one is species, and the central ring represents either species or the MIG taxonomic grouping.

## Taxonomic assignment of diverse RKN species

In addition to the *M. incognita* library, and to increase the diversity of species sequenced, we created a further three libraries of mixed species (RKN\_lib3, RKN\_lib4 and RKN\_lib5) (S1 Appendix: Section 5). This resulted in a total of six species being sequenced: j4 or immature female individuals of *M. incognita*, *M. arenaria*, *M. javanica*, *M. chitwoodi*, *M. enerolobii* and *M. hapla*, along with j2 individuals of *M. incognita*, *M. arenaria*, *M. javanica* and *M. hapla*. Additionally, to test the ability of the method on multi-species samples we created a library where each sample contained two species with a single j2 individual from each (RKN\_lib6) (S1 Appendix: Section 5). Using the optimal parameters from taxonomic assignment method development, we tested the ability of the method on these species.

## Reproducibility

A detailed protocol for DNA extraction, library preparation and sequencing is available at <http://dx.doi.org/10.17504/protocols.io.butanwie> and provided openly in the S1 Appendix of the manuscript repository available at <http://dx.doi.org/10.17605/OSF.IO/VA7S2>. The repository also includes the Kraken 2 database used in this manuscript, Kraken 2 and Recentrifuge results from our analysis, and a test dataset.

The code repository includes a detailed list of software requirements, suitable for single-step installation with the conda package manager. Bioinformatics analysis steps were

performed with the Snakemake workflow manager employing a virtual software environment built from this environment list. Raw sequence data generated in this work has been submitted to the International Nucleotide Databases under BioProject code PRJNA706653. Code repository available at [https://github.com/Graham-Sellers/RKN\\_genomic\\_taxonomic\\_assignment](https://github.com/Graham-Sellers/RKN_genomic_taxonomic_assignment).

## Results

### Taxonomic assignment method development

**Sequencing output.** Read length distributions of our test library, RKN\_lib2, are shown in Fig 1, and indicate that overall sequencing produced a mode length of close to ~ 3.5 kb. Successfully demultiplexed sequences per barcode had a mode length of ~ 4 kb (as expected for our modified library preparation protocol). Low concentration samples (barcodes 9 to 12a) had similar read length distributions compared to higher concentration samples.

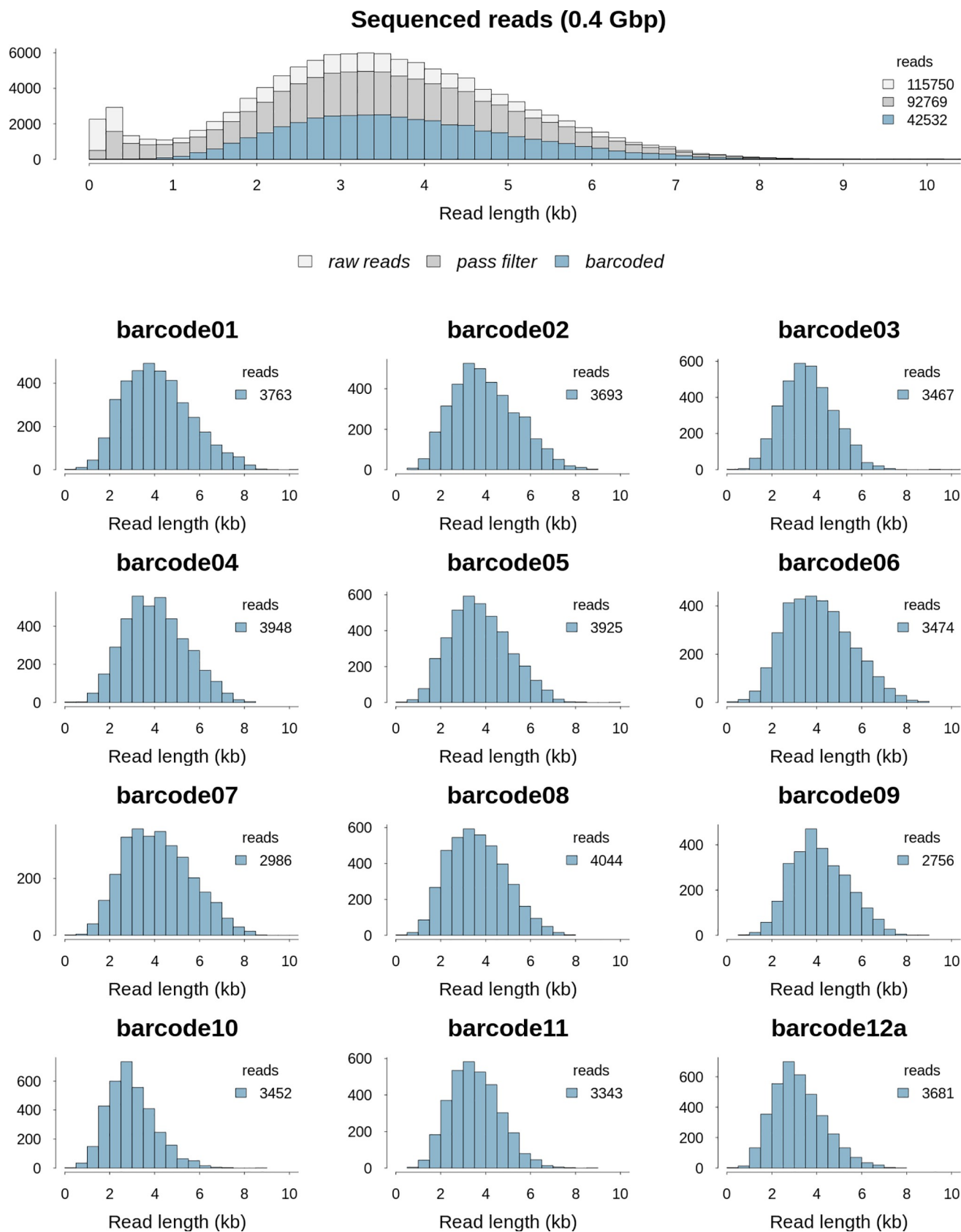
**Quality control and taxonomic assignment.** Taxonomic identification is largely very robust to the level of sequence read quality control (minimum average read quality score, Fig 2). It is clear that the classification parameters implemented by Kraken 2 have the largest effect. In particular, the most important assignment parameter is the Kraken 2 classification confidence score. However, higher values for read quality control, confidence and minimum base quality considerably reduce the number of reads assigned. To this end we determined that optimal values to reduce noise and remove incorrect assignments were a minimum average read quality score of 7 (i.e. no quality control beyond that of Guppy basecaller) and Kraken 2 confidence score of 0.02 with minimum base quality of 7.

**Read count and taxonomic assignment.** When multiplexing 12 barcoded samples per Flongle each produced between 3,000–4,000 reads per sample (Fig 1). We investigated the effect of smaller read numbers by subsampling 50 to 3000 reads and examining the proportion of reads that were correctly taxonomically assigned. Fig 3 (top panel) shows that the variation in taxonomic assignment success reduces greatly with 500 or more reads, with approximately 2.5% of reads assigned correctly to species rather than a higher level grouping. Assignment was almost exclusively to either the correct species, *M. incognita*, or to the sub-genus level MIG grouping (Fig 3, bottom panel) and this was even true for a randomised 50 read sample.

**Read length and taxonomic assignment.** Since our libraries were constructed using the ONT Rapid PCR Barcoding Kit (SQK-RPB004), few reads exceeded 6 Kb. We examined read lengths from 100 bp up to 4 kb, with each treatment having 1500 reads. Fig 4 (top panel) shows that different lengths vary between approximately 2–6% of reads assigned to the correct species (*M. incognita*), with most for the 600 bp reads. However, as shown in Fig 4 (bottom panel) the most frequent species level taxonomic assignment was to the correct species. In the case of the *M. incognita* samples, the major effect of increasing read lengths was to remove the noise of reads identified to any source other than *M. incognita* or the *M. incognita* group. Longer reads have more *k*-mers and have the effect of pulling back noise and allowing for an increase in reads classified. Reads longer than 1000 bp were decided as optimal for taxonomic assignment.

### Taxonomic assignment of diverse RKN species

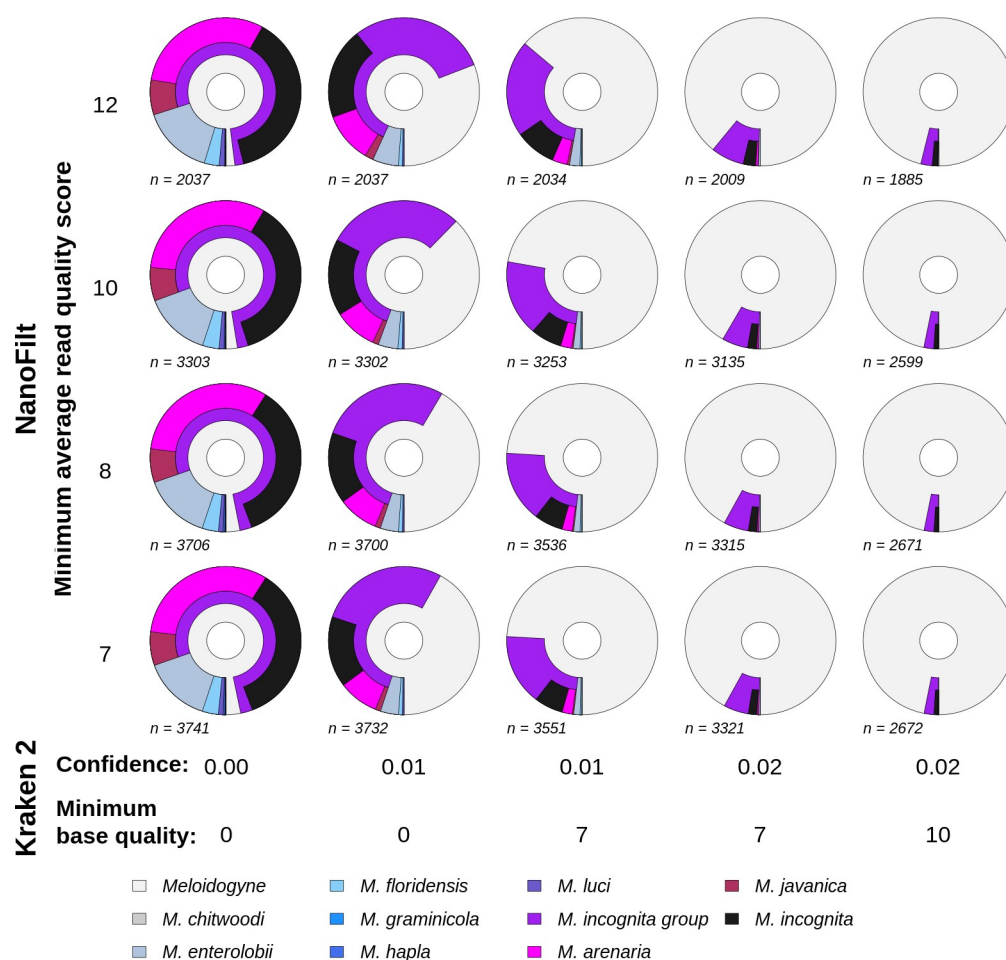
Six species of RKN were sequenced over multiple libraries; *M. incognita*, *M. arenaria*, *M. javanica*, *M. chitwoodi*, *M. enerolobii* and *M. hapla*. Within these samples were juvenile stage 2 (j2) individuals. Reads were not always evenly distributed across samples in a library. Although different nematode samples produced different read counts and qualities, in all cases, using optimal parameters (reads > 1000 bp, Kraken 2 confidence score of 0.02 and minimum base quality of 7), the most frequent assignment to the species level was the correct taxonomy



**Fig 1. Example of read length distribution from an ONT Flongle flow cell sequencing run using our method.** Top panel shows the read length distribution for the RKN\_lib2 *Meloidogyne incognita* sequencing run. Colours indicate the number of raw reads (light grey), reads to have passed filter (grey) and those demultiplexed successfully to barcoded sample (blue). Lower sections show the read length distribution for each barcoded sample. Samples are labeled with raw barcode names.

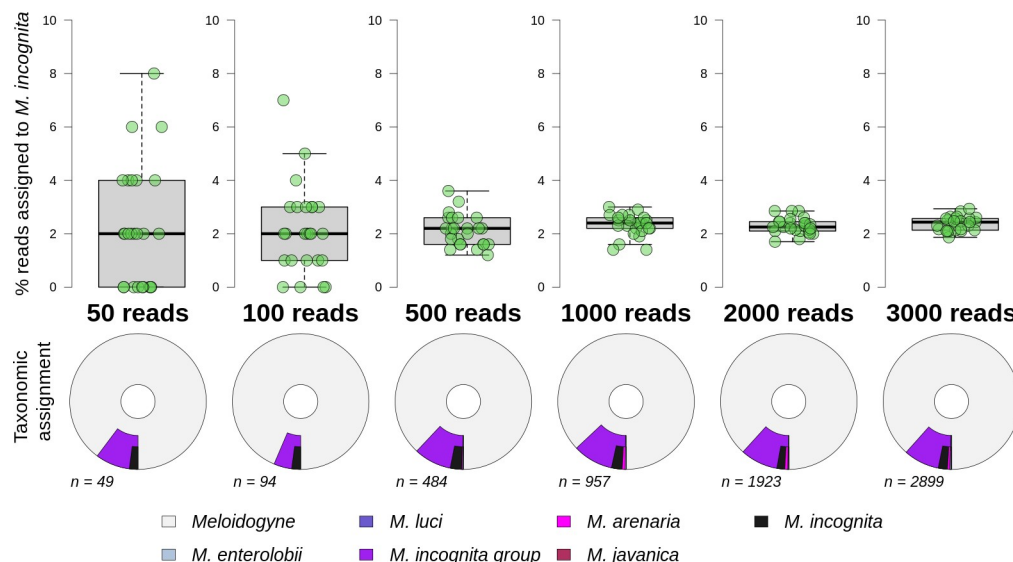
<https://doi.org/10.1371/journal.pone.0253248.g001>

(Fig 5). Species within the MIG had many reads assigned to genus level (*Meloidogyne*) and fewer to species level. In contrast, species outside the MIG (*M. chitwoodi*, *M. enterolobii* and *M. hapla*) invariably had the vast majority of reads assigned to species. Although J2 samples had no quantifiable input DNA they sequenced successfully and had taxonomic assignments similar to their j4 counterparts.



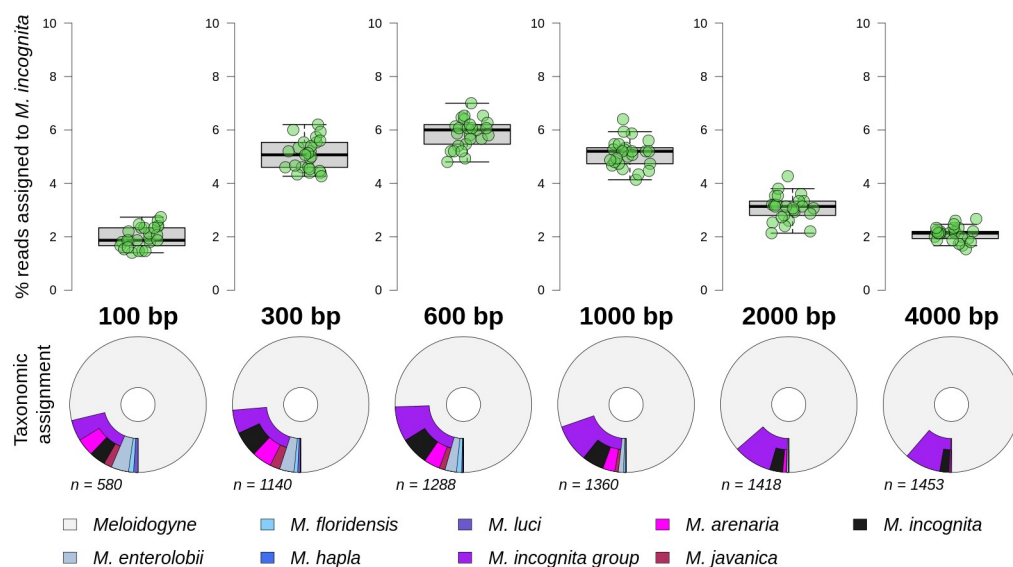
**Fig 2. Taxonomic assignment success of *M. incognita* reads varied by quality control and classification parameters.** Shown are NanoFilt minimum average read quality scores (y axis) against Kraken 2 confidence and minimum base quality parameter combinations (x axis). Only basecalled reads with a minimum average base quality of 7 were used. Kraken 2 “Confidence” indicates the minimum proportion of *k*-mers required for taxonomic assignment. Kraken 2 “Minimum base quality” indicates the base quality threshold at which *k*-mer bases are not considered ambiguous during assignment. Sections show taxonomic assignment of reads (*n*) to the genus *Meloidogyne* and lower taxonomic levels for each configuration of parameters. Each configuration is applied to the same *M. incognita* sample. For each chart the distance from the inner circle represents NCBI taxonomic lineage length; the inner segment is genus level assignment, the second segments are species level assignments (including the MIG species group). MIG species populate the outermost segments, internal to the MIG species group assignments. Segment widths denote proportion reads assigned to that taxonomic level or species.

<https://doi.org/10.1371/journal.pone.0253248.g002>



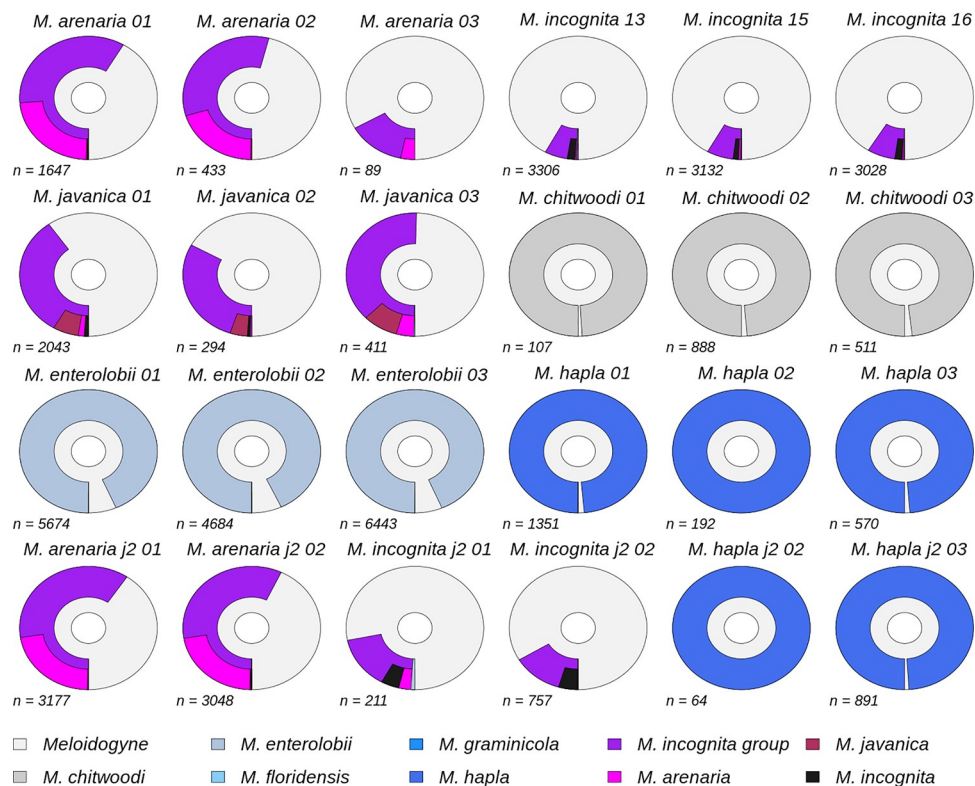
**Fig 3. Taxonomic assignment success of *M. incognita* reads varied by read count.** Top panel shows the percentage of reads assigned to species level from 50 to 3000 reads. Each treatment has 25 replicates. Lower panel shows a single replicate example of taxonomic assignment of reads ( $n$ ) to the genus *Meloidogyne* and lower taxonomic levels. For each chart the distance from the inner circle represents NCBI taxonomic lineage length; the inner segment is genus level assignment, the second segments are species level assignments (including the MIG species group). MIG species populate the outermost segments, internal to the MIG species group assignments. Segment widths denote proportion reads assigned to that taxonomic level or species.

<https://doi.org/10.1371/journal.pone.0253248.g003>



**Fig 4. Taxonomic assignment success of 1500 *M. incognita* reads varied by read length.** Top panel shows the percentage of reads assigned to species level from 100 to 4000 bp. Each treatment has 25 replicates. Lower panel shows a single replicate example of taxonomic assignment of reads ( $n$ ) to the genus *Meloidogyne* and lower taxonomic levels. For each chart the distance from the inner circle represents NCBI taxonomic lineage length; the inner segment is genus level assignment, the second segments are species level assignments (including the MIG species group). MIG species populate the outermost segments, internal to the MIG species group assignments. Segment widths denote proportion reads assigned to that taxonomic level or species.

<https://doi.org/10.1371/journal.pone.0253248.g004>



**Fig 5. Taxonomic assignment of six *Meloidogyne* species.** Top 3 rows are j4 or immature female individuals, bottom row is j2 individuals. Shown are the taxonomic assignment of reads ( $n$ ) to the genus *Meloidogyne* and lower taxonomic levels for individuals of the species. For each chart the distance from the inner circle represents NCBI taxonomic lineage length; the inner segment is genus level assignment, the second segments are species level assignments (including the MIG species group). MIG species populate the outermost segments, internal to the MIG species group assignments. Segment widths denote proportion reads assigned to that taxonomic level or species.

<https://doi.org/10.1371/journal.pone.0253248.g005>

Taxonomic assignment of multi-species samples, consisting of j2 individuals from two species (RKN\_lib6), had mixed success. Samples containing two MIG species were difficult to disentangle, whereas those with an individual external of the MIG, e.g. *M. hapla*, had clear assignments to species level for both species (see RKN\_lib6 results in the manuscript repository at <http://dx.doi.org/10.17605/OSF.IO/VA7S2>).

Under optimal circumstances for 12 multiplexed samples, with 3000 reads (assigned to *Meloidogyne* and lower) with an average read length of 4 Kb, we would expect to obtain 12 Mb per sample. This is low coverage, approximately 0.1x MIG haploid genome, and will be lower for samples that have low read counts assigned and considering PCR duplication. Despite this, we have achieved very accurate taxonomic classification of samples with low read counts (Fig 5), and therefore low genome coverage.

## Discussion

Here we demonstrated the effectiveness of a genomic approach as a method for the taxonomic assignment of a single RKN individual. Efficient DNA extraction in combination with optimised library preparation generated long-read sequencing data from ONT Flongle devices. These long-read sequences contained sufficient resolution to successfully diagnose a sample to species, even those within the closely related MIG. Genomic approaches to study RKN

typically include culturing large amounts of individuals. Here, in contrast, we show that a single j2 RKN can yield sufficient DNA for reliable identification through our method.

## Genomic approaches

Genetic characterisation of RKN has often been carried out by the PCR amplification of single or a small number of loci. These approaches can be informative but have the disadvantage that they cannot sample and represent the distinct subgenomes present in some RKN species, especially the MIG. The presence of both A and B subgenomes, shared across species, requires exceptional vigilance that homologous copies of the genome are being compared, else erroneous results could be obtained.

Genomics samples across the genome, including the A and B subgenomes of MIG. Since *k-mers* are short and exceptionally numerous they will not suffer from problems of A/B subgenome orthology but rather represent a database of oligonucleotide variants. So, unlike PCR assays, both subgenomes are sampled and analysed with no requirement for complex subgenome assignment.

We set out to develop a low-coverage sequencing approach for taxonomic ID from whole genomic DNA. The Flongle protocol that we developed returned a low genomic coverage, though this is unlikely to be an important metric when genome assembly is not involved. Instead an approach which effectively samples from the complex polyploid genome of MIG is likely more important to taxonomic ID. Our method, taking this approach, can be seen to accurately identify samples to species (Fig 5). Read counts assigned to *Meloidogyne* varied greatly between samples, but even in low read count samples the represented genome data therein had sufficient resolution for taxonomic classification. Our approach potential for samples containing more than one species. The results indicated that this might be a powerful approach for separating more genetically distinct taxa in mixed samples. It also showed promise for more closely related species (e.g. within the MIG), however its success here will almost certainly improve with the increasing completeness of the reference database.

## Scalability, cost effectiveness and time efficiency

The ONT Flongle approach developed here allows small scale analysis of up to 12 samples at a time. This is a scale that may fit well with many small laboratories. The protocol may however be scaled up considerably across multiple MinION or GridION devices with Flongle adaptors. While the ONT Flongle can generate up to 2 Gb of data, due to difficulties accessing the laboratory during the COVID-19 pandemic many of the flow cells used in our analysis were considerably past their expiration date and only produced ~ 0.5 Gb data. Despite this we were still able to produce sufficient reads for our analysis process. It is likely that future experiments or different users would yield considerably more data per barcoded sample.

The Flongle genomic approach has costs similar to other types of genetic analysis. Currently a single flow cell, reagents and consumables needed for sequencing 12 samples equates to less than 20 GBP per sample (excluding VAT and staff time) (S1 Appendix: Section 7). Our method can produce results within as little as 36 hours start to finish, with less than 4 hours expert hands-on time and a 24 hour sequencing period (S1 Appendix: Section 7).

## Taxonomic assignment of closely-related MIG species

When looking at *M. incognita*, one of the most challenging species to identify via genetic methods, many isolates are assigned not just to genus or MIG but to species, regardless of low read counts (see Fig 5). These assignments reveal the strength of that result with the percentage of reads being assigned to species and taxonomic levels.

Identification of more than one species in a single nematode preparation does not indicate hybridization or mixing of samples but rather is what would be expected with the genetic similarity of MIG species in addition to sequencing error and an imperfectly resolving database. Most reads are assigned to genus level, indicating the close relationship between species, where *k-mers* do not assign reads to species unambiguously. Despite this, we found most isolates to have a strong assignment to the correct species with a minimal percentage of reads assigned incorrectly.

Our analyses of quality control, read number, and read length indicate that taxonomic assignment of *M. incognita* is largely robust to variation in these parameters and may be accomplished on samples that have not sequenced well. We do identify that approximately 1000 reads  $\geq$  1000 bp is a powerful dataset with which to assign taxonomy in the MIG.

### Improving genomic databases improves analysis

Our study tested the identification of RKN species where the species was represented in the database. It is clear however that this will not always be the case for samples, and improving reference databases will be very advantageous to future diagnostics. We suggest expanding the public sequence databases for root-knot nematodes as a priority for our research community as a whole.

Creating genomic resources currently requires sample culturing, whole genome sequencing and assembly. In this respect, it will be interesting to see how well light-coverage, long-read sequencing from single nematodes can meaningfully supplement the genomic database. For example, if single, well-identified individuals from a species are sequenced on an ONT Flongle, would this be sufficient to include in the reference database with a minimal genome assembly method? Unfortunately the data generated from our experiments is currently not appropriate to test this, however, increasing output from Flongles and MinIONs makes this a promising future direction.

### Conclusion

We show that a low-coverage, long-read genomics approach can be used to reliably identify root-knot nematodes. Species identification was robust to much of the variation in read length and read number that we encountered. Our genomic approach is put forward as an accurate, low-cost, and scalable approach to species diagnostics. In addition, we have highlighted the importance of extending the number of agriculturally relevant species for which genomic data is in the public databases. This genomic approach has great potential to deliver more informed diagnostics of plant pathogenic nematode communities and become a valuable part of agricultural risk assessment.

### Supporting information

**S1 Appendix.** Here we provide detailed protocols for DNA extraction, library preparation, and compositions of each library used in this study.  
(PDF)

### Acknowledgments

Thanks go in particular to Dr. Jenny Spear and Prof. Dan R. Parsons for support in the development and implementation of this project. We would also like to thank Dr. Robert K. Donnelly for laboratory assistance throughout the project. We acknowledge the Viper High

Performance Computing facility of the University of Hull and its support team in achieving this work.

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## References

1. Moens M, Perry RN, Starr JL. Meloidogyne species—a diverse group of novel and important plant parasites. In: Perry R.N., Moens M., Starr J.L., editor. Root-knot nematodes. Wallingford: CAB International; 2009. pp. 1–17.
2. Trudgill DL, Blok VC. Apomictic, polyphagous root-knot nematodes: exceptionally successful and damaging biotrophic root pathogens. Annu Rev Phytopathol. 2001; 39: 53–77. <https://doi.org/10.1146/annurev.phyto.39.1.53> PMID: 11701859
3. Eves-van-den-Akker S, Jones JT. Sex: Not all that it's cracked up to be? PLoS Genet. 2018; 14: e1007160. <https://doi.org/10.1371/journal.pgen.1007160> PMID: 29470487
4. Blok VC, Powers TO. Biochemical and molecular identification. In: Perry RN, Moens M, Starr JL, editors. Root-knot nematodes. Wallingford: CAB International; 2009. pp. 119–138.
5. Karssen G, Wesemael W, Moens M. Root-knot nematodes. Plant nematology. Wallingford: CABI; 2015. pp. 73–108.
6. Álvarez-Ortega S, Brito JA, Subbotin SA. Multigene phylogeny of root-knot nematodes and molecular characterization of Meloidogyne nataliei Golden, Rose & Bird, 1981 (Nematoda: Tylenchida). Sci Rep. 2019; 9: 11788. <https://doi.org/10.1038/s41598-019-48195-0> PMID: 31409860
7. Triantaphyllou AC. Cytogenetics, cytotaxonomy and phylogeny of root-knot nematodes. In: Sasser CCC J. N., editor. An advanced treatise on Meloidogyne, Vol I Biology and control. Raleigh: North Carolina State University Graphics; 1985. pp. 113–126.
8. Dalmaso A, Berge JB. Enzyme polymorphism and the concept of parthenogenetic species, exemplified by Meloidogyne. Concepts in nematode systematics. 1983; 187–196.
9. Abad P, Gouzy J, Aury J-M, Castagnone-Sereno P, Danchin EGJ, Deleury E, et al. Genome sequence of the metazoan plant-parasitic nematode Meloidogyne incognita. Nat Biotechnol. 2008; 26: 909–915. <https://doi.org/10.1038/nbt.1482> PMID: 18660804
10. Blanc-Mathieu R, Perfus-Barbeoch L, Aury J-M, Da Rocha M, Gouzy J, Sallet E, et al. Hybridization and polyploidy enable genomic plasticity without sex in the most devastating plant-parasitic nematodes. PLoS Genet. 2017; 13: e1006777. <https://doi.org/10.1371/journal.pgen.1006777> PMID: 28594822
11. Szitenberg A, Salazar-Jaramillo L, Blok VC, Laetsch DR, Joseph S, Williamson VM, et al. Comparative Genomics of Apomictic Root-Knot Nematodes: Hybridization, Ploidy, and Dynamic Genome Change. Genome Biol Evol. 2017; 9: 2844–2861. <https://doi.org/10.1093/gbe/evx201> PMID: 29036290
12. Koutsovoulos GD, Pouillet M, Elashry A, Kozłowski DKL, Sallet E, Da Rocha M, et al. Genome assembly and annotation of Meloidogyne enterolobii, an emerging parthenogenetic root-knot nematode. Sci Data. 2020; 7: 324. <https://doi.org/10.1038/s41597-020-00666-0> PMID: 33020495
13. Susič N, Koutsovoulos GD, Riccio C, Danchin EGJ, Blaxter ML, Lunt DH, et al. Genome sequence of the root-knot nematode Meloidogyne luci. J Nematol. 2020; 52: 1–5. <https://doi.org/10.21307/jofnem-2020-025> PMID: 32180388

14. Wood DE, Lu J, Langmead B. Improved metagenomic analysis with Kraken 2. *Genome Biol.* 2019; 20: 257. <https://doi.org/10.1186/s13059-019-1891-0> PMID: 31779668
15. Sellers GS, Di Muri C, Gómez A, Hänfling B. Mu-DNA: a modular universal DNA extraction method adaptable for a wide range of sample types. *MBMG.* 2018; 2: e24556.
16. Rohland N, Reich D. Cost-effective, high-throughput DNA sequencing libraries for multiplexed target capture. *Genome Res.* 2012; 22: 939–946. <https://doi.org/10.1101/gr.128124.111> PMID: 22267522
17. Sato K, Kadota Y, Gan P, Bino T, Uehara T, Yamaguchi K, et al. High-Quality Genome Sequence of the Root-Knot Nematode *Meloidogyne arenaria* Genotype A2-O. *Genome Announc.* 2018;6. <https://doi.org/10.1128/genomeA.00519-18> PMID: 29954888
18. Lunt DH, Kumar S, Koutsovoulos G, Blaxter ML. The complex hybrid origins of the root knot nematodes revealed through comparative genomics. *PeerJ.* 2014; 2: e356. <https://doi.org/10.7717/peerj.356> PMID: 24860695
19. Phan NT, Orjuela J, Danchin EGJ, Klopp C, Perfus-Barbeoch L, Kozłowski DK, et al. Genome structure and content of the rice root-knot nematode (*Meloidogyne graminicola*). *Ecol Evol.* 2020; 13: 626. <https://doi.org/10.1002/ece3.6680> PMID: 33144944
20. Somvanshi VS, Tathode M, Shukla RN, Rao U. Nematode Genome Announcement: A Draft Genome for Rice Root-Knot Nematode, *Meloidogyne graminicola*. *J Nematol.* 2018; 50: 111–116. <https://doi.org/10.21307/jofnem-2018-018> PMID: 30451432
21. Opperman CH, Bird DM, Williamson VM, Rokhsar DS, Burke M, Cohn J, et al. Sequence and genetic map of *Meloidogyne hapla*: A compact nematode genome for plant parasitism. *Proc Natl Acad Sci U S A.* 2008; 105: 14802–14807. <https://doi.org/10.1073/pnas.0805946105> PMID: 18809916
22. Asamizu E, Shirasawa K, Hirakawa H, Iwahori H. Root-knot nematode genetic diversity associated with host compatibility to sweetpotato cultivars. *Mol Plant Pathol.* 2020; 21: 1088–1098. <https://doi.org/10.1111/mpp.12961> PMID: 32558191
23. Abad P, Gouzy J, Aury J-M, Castagnone-Sereno P, Danchin EGJ, Deleury E, et al. Genome sequence of the metazoan plant-parasitic nematode *Meloidogyne incognita*. *Nat Biotechnol.* 2008; 26: 909–915. <https://doi.org/10.1038/nbt.1482> PMID: 18660804
24. Köster J, Rahmann S. Snakemake—a scalable bioinformatics workflow engine. *Bioinformatics.* 2012; 28: 2520–2522. <https://doi.org/10.1093/bioinformatics/bts480> PMID: 22908215
25. De Coster W, D'Hert S, Schultz DT, Cruts M, Van Broeckhoven C. NanoPack: visualizing and processing long-read sequencing data. *Bioinformatics.* 2018; 34: 2666–2669. <https://doi.org/10.1093/bioinformatics/bty149> PMID: 29547981
26. Martí JM. Recentrifuge: Robust comparative analysis and contamination removal for metagenomics. *PLoS Comput Biol.* 2019; 15: e1006967. <https://doi.org/10.1371/journal.pcbi.1006967> PMID: 30958827
27. R Core Team. R: A Language and Environment for Statistical Computing. Vienna, Austria: R Foundation for Statistical Computing; 2021. Available: <https://www.r-project.org/>
28. Gu Z, Gu L, Eils R, Schlesner M, Brors B. circlize Implements and enhances circular visualization in R. *Bioinformatics.* 2014; 30: 2811–2812. <https://doi.org/10.1093/bioinformatics/btu393> PMID: 24930139