mlPEA User Manual

(version 1.0)

- mlPEA is a user-friendly and multi-functionality platform specifically tailored to the needs of streamlined processing of m^{6}A-Seq data in a reference genome-free manner. By taking advantage of machine learning (ML) algorithms, mlPEA enhanced the m^{6}A-Seq data analysis by constructing robust computational models for identifying high-quality transcripts and high-confidence m^{6}A-modified regions.
- mlPEA comprises four functional modules: **Data Preprocessing, Transcriptome Construction, m^{6}A Calling, and Functional Exploration**.
- mlPEA was powered with an advanced packaging technology, which enables compatibility and portability.
- mlPEA project is hosted on http://github.com/cma2015/mlPEA
- mlPEA docker image is available at http://hub.docker.com/r/malab/mlpea

Transcriptome Construction Module

This module provides step-by-step functions required for transcriptome construction.

Transcriptome Assembly

Four assemblers that support strand-specific m^{6}A-Seq data are wrapped to use. Currently, **Trinity**, **rnaSPAdes**, **TransABySS**, **TransLiG**.

Tools	Description	Input	Output	Time (test data)	Reference
Trinity	Trinity partitions the sequence data into many individual de Bruijn graphs, each representing the transcriptional complexity at a given gene or locus, and then processes each graph independently to extract full-length splicing isoforms and to tease apart transcripts derived from paralogous genes.	Sequencing reads in FASTQ format	Transcripts in FASTA format	~4 mins	(Grabherr *et al.*, 2011)
rnaSPAdes	rnaSPAdes has been developed on top of the SPAdes genome assembler and explores computational parallels between assembly of transcriptomes and single-cell genomes.	Sequencing reads in FASTQ format	Transcripts in FASTA format	~3 mins	(Prjibelski *et al.*, 2020)
TransABySS	A *de novo* short-read transcriptome assembly and analysis pipeline that addresses variation in local read densities by assembling read substrings with varying stringencies and then merging the resulting contigs before analysis.	Sequencing reads in FASTQ format	Transcripts in FASTA format	~5 mins	(Robertson *et al.*, 2010)
TransLiG	TransLiG is shown to be significantly superior to all the salient *de novo* assemblers in both accuracy and computing resources when tested on artificial and real RNA-seq data.	Sequencing reads in FASTQ format	Transcripts in FASTA format	~4 mins	(Liu *et al.*, 2019)

Transcriptome Screening

Transcriptome screening implements three pipelines for transcripts deduplication, respectively.

Tools	Description	Input	Output	Time (test data)	Reference
Merge and Deduplication	A novel program CD-HIT for clustering biological sequences to reduce sequence redundancy and improve the performance of other sequence analyses.	Multiple Transcripts files in FASTA format	Transcripts in FASTA format	~5 mins	(Fu *et al*., 2012)
Reads Coverage-based Screening	An ultrafast and memory-efficient tool Bowtie 2 for aligning sequencing reads, reads coverage was measured by SAMtools and BEDTools.	Transcripts in FASTA format	Transcripts in FASTA format	~20 mins	(Langmead *et al*., 2009; Langmead *et al*., 2012) (Danecek *et al*., 2021) (Quinlan *et al*., 2010)
Machine Learning-based Screening	An ML-based classification model to distinguish high-quality transcripts from noisy ones using the random forest-based PSoL algorithm, which requires only a set of positive samples. The positive sample set is constructed by clustering analysis of assembled transcripts and already annotated mRNAs using a fast and sensitive sequence alignment method MMseqs2. It provides more than 670 features to encode each transcript sequence.	Transcripts in FASTA format	Transcripts in FASTA format	~10mins	(Steinegger *et al*., 2017) (Wang *et al*., 2023) (Wang *et al*., 2006)

Transcriptome Assembly

Currently, mlPEA wrapped four assemblers that support strand-specific m^{6}A-Seq data to *de nove* assemble transcripts. Here, we take Trinity as an example to show how to use mlPEA to run transcriptome assembly, the other three assemblers are similar.

Input

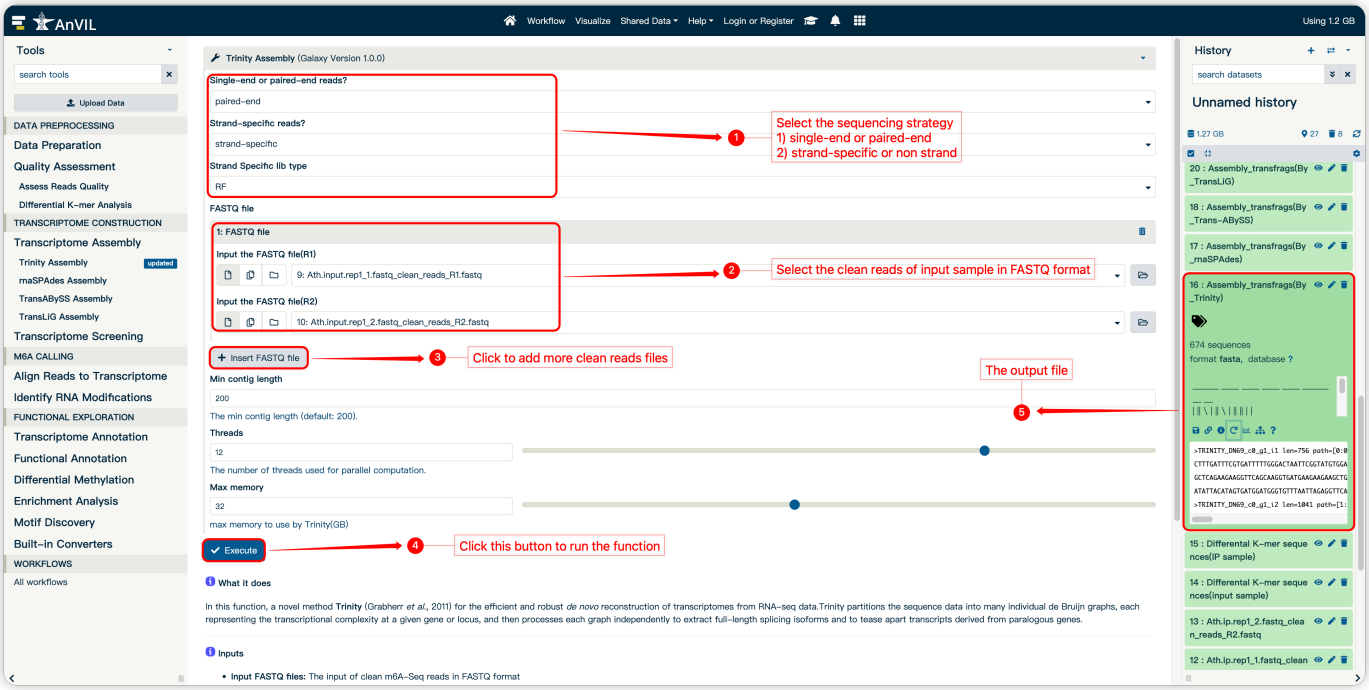
- **Input FASTQ files:** cleaned m^{6}A-Seq input reads in FASTQ format

Output

- ***De novo* assembled transcripts by Trinity in FASTA format**

How to use this function

- The following screenshot shows us how to use this function



Merge and Deduplication

In this function, a novel program **CD-HIT** for clustering biological sequences to reduce sequence redundancy and improve the performance of other sequence analyses.

Input

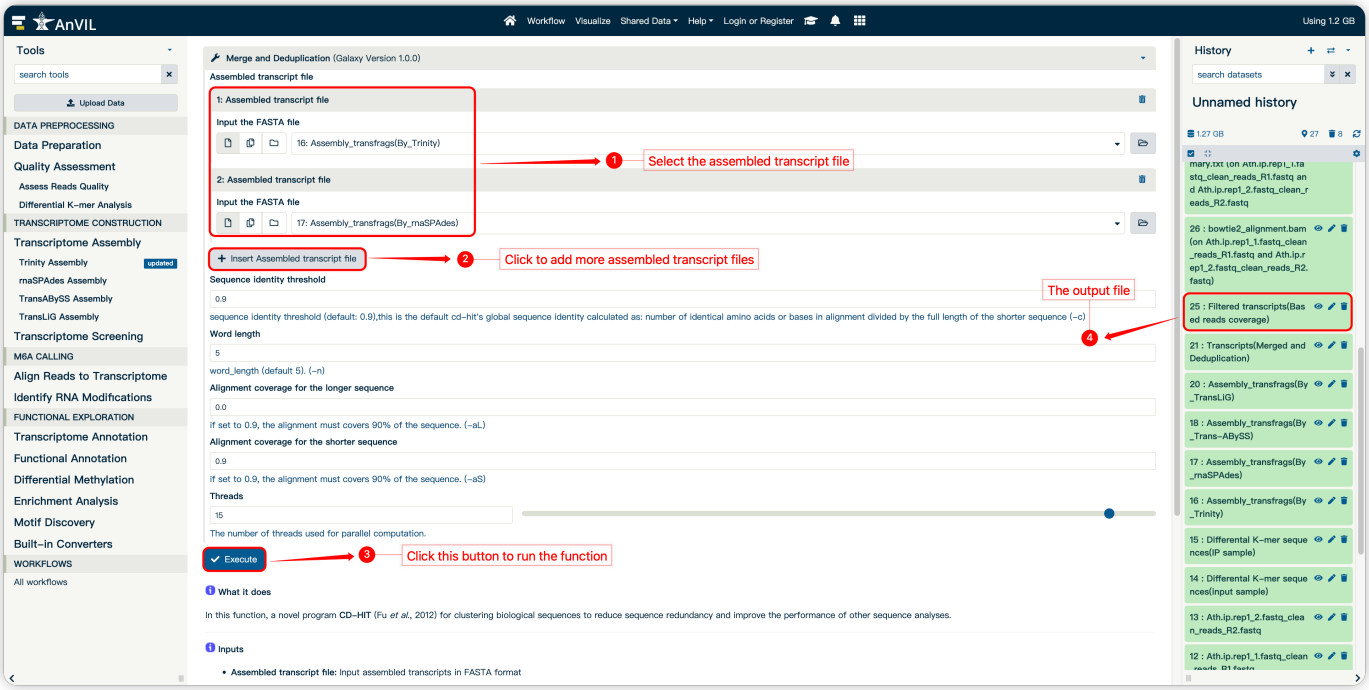
- **Input FASTA files:** assembled transcripts in FASTA format

Output

- **Transcripts after deduplication in FASTA format**

How to use this function

- The following screenshot shows us how to use this function



Reads Coverage-based Screening

In this function, we developed a pipline for screening transcripts based on reads coverage. **Bowtie2** for aligning sequencing reads, reads coverage was measured by **SAMtools** and **BEDTools**.

Input

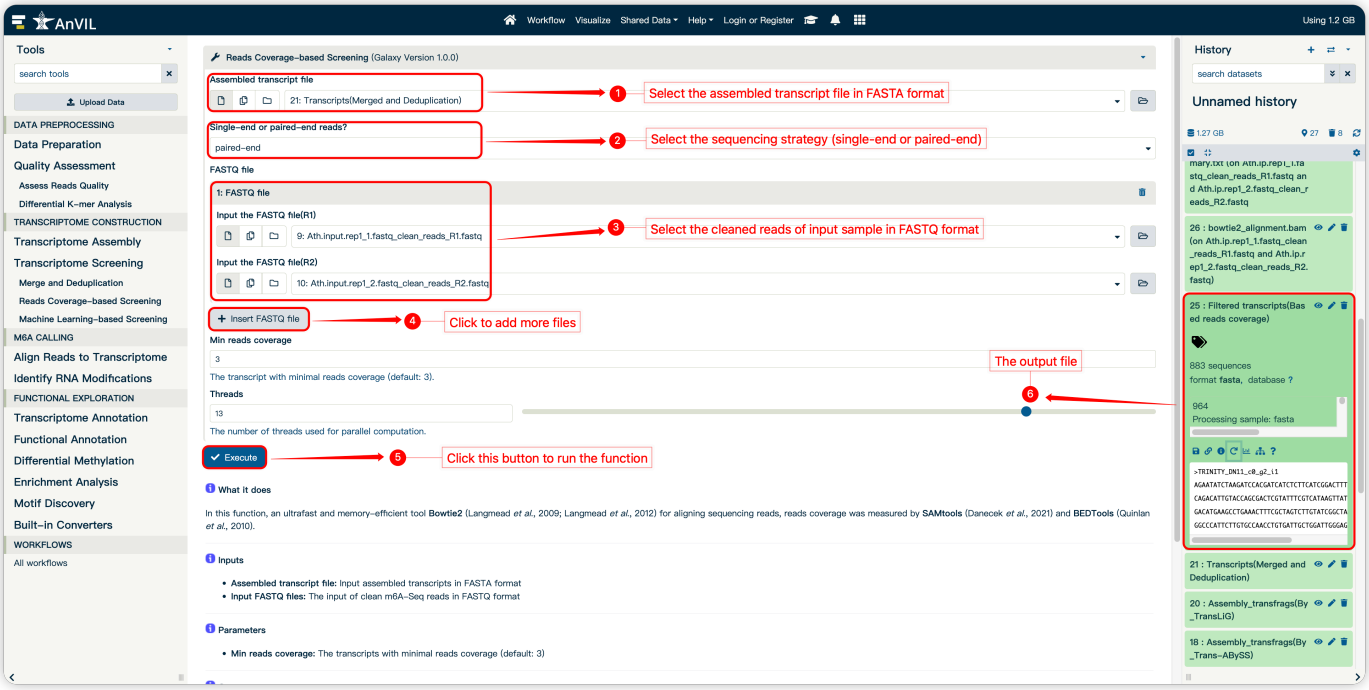
- **Input FASTQ files:** cleaned m^{6}A-Seq input reads in FASTQ format

Output

- ***De novo* assembled transcripts in FASTA format**

How to use this function

- The following screenshot shows us how to use this function



Machine Learning-based Screening

In this function, an ML-based classification model to distinguish high-quality transcripts from noisy ones using the random forest-based Positive Sample only Learning (PSoL) algorithm, which requires only a set of positive samples. The positive sample set is constructed by clustering analysis of assembled transcripts and already annotated mRNAs using a fast and sensitive sequence alignment method **MMseqs2**. It provides more than 670 features to encode each transcript sequence, including 177 sequence-intrinsic features, 399 physico-chemical features and 101 structure-based features.

Input

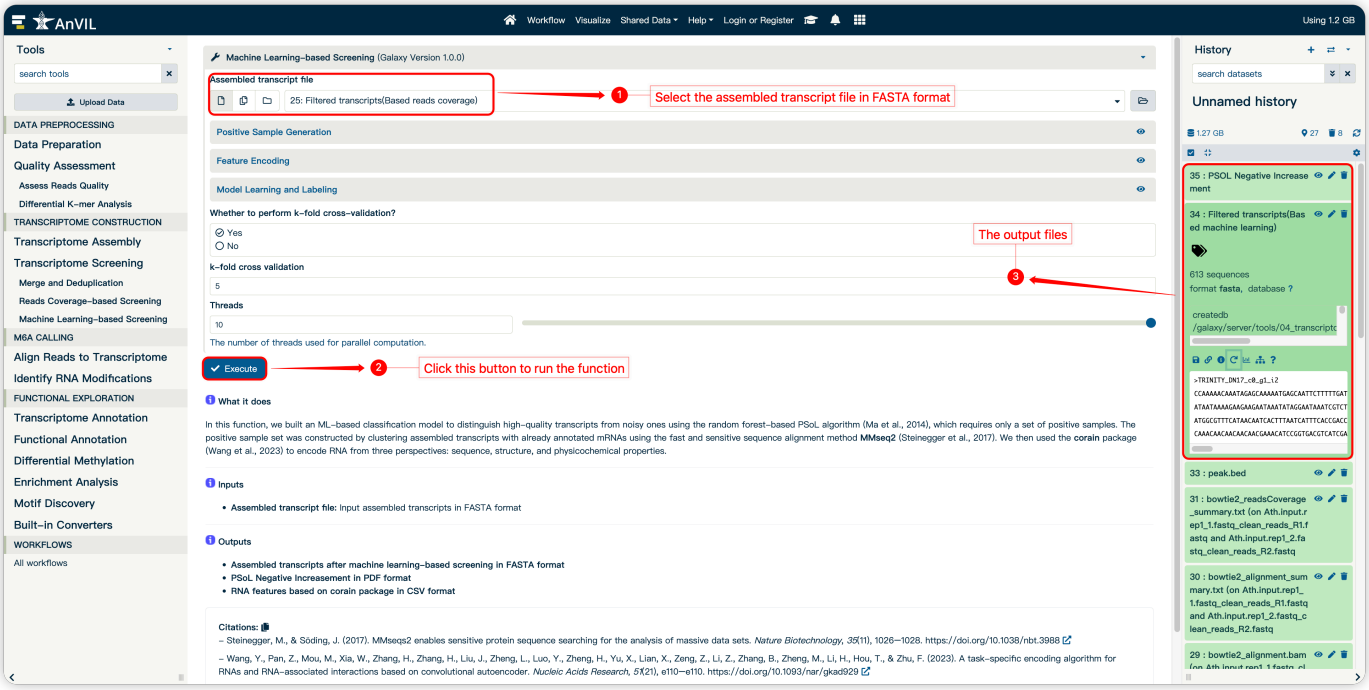
- **Input FASTQ files:** *de novo* assembled transcripts in FASTA format

Output

- **High-quality transcripts in FASTA format**
- **PSoL Negative Increasement in PDF format**
- **RNA features based on corain package in CSV format**

How to use this function

- The following screenshot shows us how to use this function


AnVIL
Workflow
Visualize
Shared Data
Help
Login or Register
Using 1.2 GB
Tools
search tools
Upload Data
DATA PREPROCESSING
Data Preparation
Quality Assessment
Assess Reads Quality
Differential K-mer Analysis
TRANSCRIPTOME CONSTRUCTION
Transcriptome Assembly
Transcriptome Screening
Merge and Deduplication
Reads Coverage-based Screening
Machine Learning-based Screening
M6A CALLING
Align Reads to Transcriptome
Identify RNA Modifications
FUNCTIONAL EXPLORATION
Transcriptome Annotation
Functional Annotation
Differential Methylation
Enrichment Analysis
Motif Discovery
Built-in Converters
WORKFLOWS
All workflows
Machine Learning-based Screening (Galaxy Version 1.0.0)
Assembled transcript file
25: Filtered transcripts(Based reads coverage)
1
Select the assembled transcript file in FASTA format
Positive Sample Generation
Feature Encoding
Model Learning and Labeling
Whether to perform k-fold cross-validation?
Yes
No
The output files
3
k-fold cross validation
5
Threads
10
The number of threads used for parallel computation.
Execute
2
Click this button to run the function
What it does
In this function, we built an ML-based classification model to distinguish high-quality transcripts from noisy ones using the random forest-based PSoL algorithm (Ma et al., 2014), which requires only a set of positive samples. The positive sample set was constructed by clustering assembled transcripts with already annotated mRNAs using the fast and sensitive sequence alignment method MMseq2 (Steinegger et al., 2017). We then used the corain package (Wang et al., 2023) to encode RNA from three perspectives: sequence, structure, and physicochemical properties.
Inputs
Assembled transcript file: Input assembled transcripts in FASTA format
Outputs
Assembled transcripts after machine learning-based screening in FASTA format
PSoL Negative Increasement in PDF format
RNA features based on corain package in CSV format
Citations:
– Steinegger, M., & Söding, J. (2017). MMseqs2 enables sensitive protein sequence searching for the analysis of massive data sets. Nature Biotechnology, 35(11), 1026–1028. https://doi.org/10.1038/nbt.3988
– Wang, Y., Pan, Z., Mou, M., Xia, W., Zhang, H., Zhang, H., Liu, J., Zheng, L., Luo, Y., Zheng, H., Yu, X., Lian, X., Zeng, Z., Li, Z., Zhang, B., Zheng, M., Li, H., Hou, T., & Zhu, F. (2023). A task-specific encoding algorithm for RNAs and RNA-associated interactions based on convolutional autoencoder. Nucleic Acids Research, 51(21), e110–e110. https://doi.org/10.1093/nar/gkad929
History
search datasets
Unnamed history
1.27 GB
27
8
35 : PSOL Negative Increasement
34 : Filtered transcripts(Based machine learning)
613 sequences
format fasta, database ?
createdb /galaxy/server/tools/04_transcriptc
>TRINITY_DN17_c0_g1_i2
CCAAAAACAAATAGAGCAAAAATGAGCAATTCTTTTTGAT
ATAATAAAAGAAGAAGAATAAATATAGGAATAAATCGTCT
ATGGCGTTTCATAACAATCACTTTAATCATTTCACCGACC
CAAACAACACAACAACGAAACATCCGGTGACGTCATCGA
33 : peak.bed
31 : bowtie2_readsCoverage_summary.txt (on Ath.input.rep1_1.fastq_clean_reads_R1.fastq and Ath.input.rep1_2.fastq_clean_reads_R2.fastq
30 : bowtie2_alignment_summary.txt (on Ath.input.rep1_1.fastq_clean_reads_R1.fastq and Ath.input.rep1_2.fastq_clean_reads_R2.fastq
29 : bowtie2_alignment.bam