\usepackage{booktabs} \usepackage{longtable} \usepackage{array}
\usepackage{multirow} \usepackage{wrapfig} \usepackage{float}
\usepackage{colortbl} \usepackage{pdflscape} \usepackage{tabu}
\usepackage{threeparttable} \usepackage{threeparttablex}
\usepackage[normalem]{ulem} \usepackage{makecell} \usepackage{xcolor}

Transcriptome Data Analysis in Nonmodel Organisms

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3/10/23

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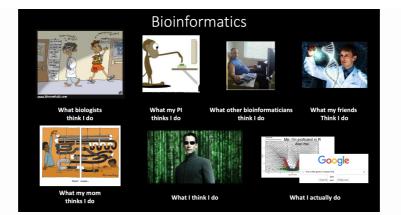
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Preface

Welcome to the book of Transcriptome Data Analysis in Non-Model Organisms.

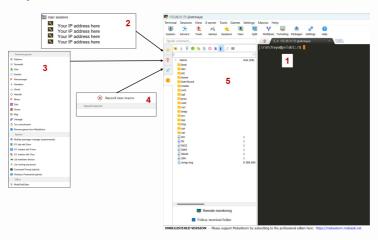


A confusion matrix. From twitter <u>@ninadoak</u>.

Introduction to MobaXterm, Terminal, and SSH

MobaXterm (for Windows)

MobaXterm is a toolbox for remote computing. In a single Windows application, it provides loads of functions that are tailored for programmers, webmasters, IT administrators and pretty much all users who need to handle their remote jobs in a more simple fashion. MobaXterm provides all the important remote network tools, such as SSH, X11, RDP, VNC, FTP, MOSH, and of course, Unix commands, and many more!



MobaXterm user interface. In the context of remote access through SSH and FTP, mobaXterm provides easy-to-access route as (1) a secure shell (SSH) terminal of the remote server, (2) a list of remote server you've accessed, (3) Utilities facilitating remote server access including entertainment, like Swiss army knife!, (4) If you want to reduce redundant typing, just set macro to it, and (5) a files available in the current working directory in the remote server, you can also transfer files from remote server to your local computer using this route! There are many advantages of having an All-In-One network application for your remote tasks, e.g. when you use SSH to connect to a remote server, a graphical SFTP browser will automatically pop up in order to directly edit your remote files.

Visit MobaXterm official website to see a demo: https://mobaxterm.mobatek.net/demo.html

Terminal (for macOS)

Terminal provides a command-line interface to macOS. Each window in Terminal represents an instance of a shell process. The window contains a prompt that indicates you can enter a command. The prompt you see depends on your Terminal and shell settings, but it often includes the name of the host you're logged in to, your current working folder, your user name, and a prompt symbol. For example, if a user named michael is using the default zsh shell, the prompt appears as:

```
michael@MacBook-Pro ~ %
```

This indicates that the user named michael is logged in to a computer named MacBook-Pro, and the current folder is his home folder, indicated by the tilde (~).

MacOS features a built-in SSH client called Terminal which allows you to quickly and easily connect to a server. Starting from open the "terminal" app, and enter the standard SSH command:

```
ssh user@IPAddress
```

This will connect to the server via SSH with the username `user` and the default SSH port 22. The connection will look similar to the following:

Connecting to Remote Server

Bioinformatics data processing tasks require more computing power than our laptops, so we need large servers or clusters. It's likely you'll work mostly over a network connection with remote machines on some projects. It can be frustrating for beginners to work with a remote machine. So, This part will introduce you to some commonly used bash commands. To make it easier for beginners to manage their remote machines, there are a range of different tools and technologies available, such as SSH, FTP, and terminal commands, which can be used to access and manage the environment of the machine. Additionally, there are a variety of bash commands which can be used to streamline the process of managing the machine Buffalo (2015).

What you need to know for connecting to a remove server:

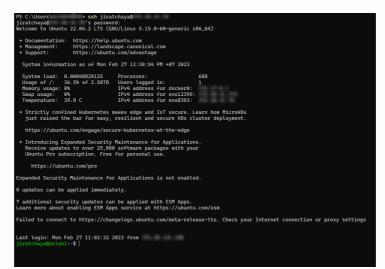
- 1. Your username and password in the remote server
- 2. IP address of the remote server, and which port to connect to server
- 3. You should know whether the remote server accessible via local network or a public IP address

By default, SSH uses port 22 but it can be changed to a non-standard port. To securely connect the client to the remote server, SSH uses symmetric encryption, asymmetric encryption, and hashing. If you're connecting for the first time, you'll be asked to verify the server's public key. Whenever you connect to the same server in the future, the client will reference this verified public key. During an SSH connection, the client and server negotiate a session key used to encrypt and decrypt data.



In order to establish a connection, SSH needs to verify SHA keys once connected for the first time. Once authentication is complete, the SSH connection is secure and can be trusted for future access.

Upon connecting to the remote server, you'll see a welcome message like this



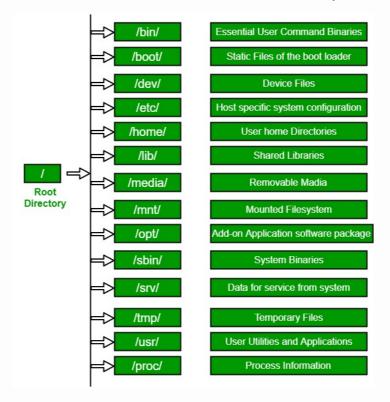
An example welcome message of server using Ubuntu, including general software and hardware status, information of the latest connection, as well as a prompt for user command.

Bash Command Language for Biologists

Shell scripts (or shell or UNIX) are widely used in bioinformatics because they're the interface for large bioinformatics programs. In this workshop, you'll learn how to use the necessary Bash command concepts. This will allow you to focus on the content of the commands in the following chapters rather than on understanding shell syntax. However, before we start learning bash, it's good to understand Linux file systems a little bit.

Linux File Systems

In Unix-like operating systems, the Linux file system defines the directory structure and contents. Even if they're located on different physical or virtual hard disks, all files and directories are located under the root directory.



Schematic hierarchy of Linux file systems. The figure is adopted from <u>https://www.geeksforgeeks.org/linux-file-hierarchy-structure</u>.

Root (/)

It is the root directory of the entire file system hierarchy and the primary hierarchy root. The root directory is where everything begins. This directory can be written only by root.

/bin

Essential commands that must be available with all users, for example, cat, ls, cp, cd, top, mkdir and many more.

/dev

Essential device files such as /dev/null, /dev/shm. This includes terminal devices, USB or other devices connected to the system.

/etc

System-wide configuration files for the host, contain files that all programs need. Also included are startup and shutdown shell scripts for starting and stopping individual programs, such as /etc/fstab for permanently mounting external disks, /etc/netplan for configuring the network and IP address, and more.

/home

Users' home directories, where they keep their saved files and settings. These directories are used to store all of a user's files and settings in one place so that they can easily access their data and keep it organized. For example /home/ponsit, /home/jiratchaya, /home/prasert.

/lib

Contain essential libraries for the binaries in /bin/ and /sbin/.

/media

Mount points for removable media such as CD-ROMs (deprecated).

/mnt

Temporary mount directory where sysadmins can mount file systems, such as /mnt/external_disk_1, /mnt/removable_drive_1, etc.

/opt

Optional application software packages, including add-on applications from individual vendors.

/sbin

Essential system binaries, e.g., fsck, init, route.

/tmp

Temporary files that aren't preserved between reboots and may be severely restricted.

/usr

A secondary hierarchy for read-only user data. Most utilities and applications are located here.

Basic Bash Commands

Bash is a Unix shell that allows you to enter commands that are then interpreted and executed by the computer. Commands can be used to perform tasks such as creating a directory, running a program, or deleting a file. Bash is a type of interpreter that takes user input and converts it into a language that the computer can understand and execute. Commands usually consist of keywords, arguments, and flags that allow the user to control how the command is interpreted and executed by the computer.

Creating directories

Keeping all your files in a single directory makes things much easier for you and your collaborators, and makes it easier to reproduce. Suppose you're working on a transcriptome analysis of *Cyanophora paradoxa*. Your first step would be to choose a short, appropriate project name and create some basic directories.

Note: In Linux file system, **directory** is exactly the same with **folder**.

To keep it short and clear, 'Cpa' is used as an alias article name for *C. paradoxa*, and as the name of the directory, followed by words describing your work, for example.

□ warning: Avoid using spaces () or special characters such as slashes (/), backslashes (\), accented characters ('), tilde (~), and many others. It is **recommended to use underscore (_)** or hyphen (-) instead of these special characters.

• Create a directory name 'Cpa_RNASeq' from current working directory

mkdir Cpa_RNASeq

This will create a directory named 'Cpa_RNASeq' in your current working directory. Let us create some subdirectories!

• Create subdirectory '01_Rawdata' under the 'Cpa_RNASeq' directory

mkdir Cpa_RNASeq/01_Rawdata

This will create a subdirectory name '01_Rawdata' in the directory 'Cpa_RNASeq'

• Create multiple directory at once

For example, if you want to create 2 directories named '02_QC' and '03_assembly' under the 'Cpa_RNASeq' directory, then simply type

```
mkdir Cpa_RNASeq/{02_QC,03_assembly}
```

Activity

A well-organized project directory can make life easier. Your project directory should be organized in a consistent and understandable way. A clear project organization makes it easier for both you and your collaborators to find out exactly where and what everything is located, which improves the reproducibility of research. It's also much easier to automate tasks when files are organized and clearly named.

In this workshop, we'll learn transcriptome data analysis in many steps from downloading reads to transcriptome annotation. Therefore, we'll divide each analysis step into subdirectories as follows. Let's assume that we have already created the directory Cpa_RNASeq.

└── Cpa_RNASeq

```
├── 01_Rawdata
```

```
└── 02_QC
```

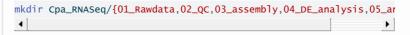
```
├── 03_assembly
```

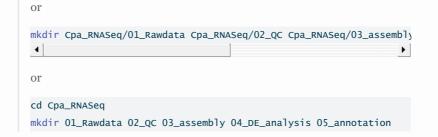
```
├-- 04_DE_analysis
```

```
└── 05_annotation
```

Could you generate bash command(s) to create these directories ?

Answer





Navigating your file system

The file system manages the files and directories of the operating system. It organizes our data into files, which store information, and directories. When you see the prompt below on your terminal's screen, it means that your terminal has processed the command you entered and is ready for the next command.

```
jiratchaya@DESKTOP-P2DD13C:~$
```

jiratchaya is username using this terminal. The address @ symbol followed by DESKTOP-P2DD13C in a computer or server name. And, the dollar sign \$ is a prompt, which shows us that the shell is waiting for input. Your shell may use a different character as a prompt and may add information before the prompt.

If you want to find out where we are now, type

pwd

pwd stands for print working directory. Without explicit specification, the computer assumes that we want to execute commands in our current working directory. This can be a user's home directory (~).

If you want to change the directory, e.g. to the 'Cpa_RNASeq' directory we just created, just type the following

cd Cpa_RNASeq

cd stands for "change directory". You can change our working directory by typing cd followed by a directory name. In this case you change from the current directory to the directory named 'Cpa_RNASeq'.

Listing directories

We can see what files and subdirectories are in this directory by running ls, which stands for "listing":

٦s

Expected result:

```
jiratchaya@DESKTOP-P2DD13C:~/Cpa_RNASeq$ ls
01_Rawdata 02_QC 03_assembly
```

Let us look at the other way. This way is to list all the files and directories, including the users who own them, the permissions, and the file size in bytes.

ls -1

Expected result:

```
jiratchaya@DESKTOP-P2DD13C:~/Cpa_RNASeq$ ls -1
total 12
drwxr-xr-x 2 jiratchaya jiratchaya 4096 Mar 1 21:02 01_Rawdata
drwxr-xr-x 2 jiratchaya jiratchaya 4096 Mar 1 21:02 02_QC
drwxr-xr-x 2 jiratchaya jiratchaya 4096 Mar 1 21:02 03_assembly
```

List files and folders, permissions and size in a human readable format.

ls -1h

Expected result:

```
jiratchaya@DESKTOP-P2DD13C:~/Cpa_RNASeq$ ls -1
total 12
drwxr-xr-x 2 jiratchaya jiratchaya 4096 Mar 1 21:02 01_Rawdata
drwxr-xr-x 2 jiratchaya jiratchaya 4096 Mar 1 21:02 02_QC
drwxr-xr-x 2 jiratchaya jiratchaya 4096 Mar 1 21:02 03_assembly
```

See all hidden files and directories

ls -la

Expected result:

```
jiratchaya@DESKTOP-P2DD13C:~/Cpa_RNASeq$ ls -la
total 20
drwxr-xr-x 5 jiratchaya jiratchaya 4096 Mar 1 21:02 .
drwxr-xr-x 2 jiratchaya jiratchaya 4096 Mar 1 21:02 01_Rawdata
drwxr-xr-x 2 jiratchaya jiratchaya 4096 Mar 1 21:02 01_Rawdata
drwxr-xr-x 2 jiratchaya jiratchaya 4096 Mar 1 21:02 02_QC
```

Files and directories handling

Creating and editing files

When you work on the command line, you often need to create or edit text files. In this workshop, we recommend using nano as a text editor. Other Unix text editors you may have heard of include vi, vim, emacs, vscode, and many more.

We'll create the file test.fasta. To open an existing file or create a new file, type nano followed by the filename:

nano test.fasta

This will bring up the text editing screen on your terminal. Here you can type anything you want, but in this case we'll create a sequence like this.

>seq_01 TCGCTAGTC

>seq_02 TAGCGAGTT □**Note:**Always leave an enter in the last line. This is advantageous if this file is further used by many programmes.



The text editing screen is displayed once you have typed nano into some files. At the bottom of the window is a list of the most important keyboard shortcuts for the nano editor. All commands are preceded by either a ^ or an M character. The caret symbol (^) stands for the Ctrl key. For example, the commands ^J mean that you press the Ctrl and J keys simultaneously. The letter M stands for the Alt key.

To edit a file, you can use the navigation keys such as arrow keys, End, Home, PgUp or PgDn to control the cursor.

To save the changes you made to the file, press ctrl+o. If the file doesn't exist yet, it'll be created after saving.

To exit nano, press ctrl+x. If there are unsaved changes, you'll be asked if you want to save the changes. Nano will ask you 'Save modified buffer?', then type y to confirm the edit.

Copying files and directories

To copy files and directories the command cp can be used. cp stands for copy and is used to copy files and directories in Linux. An example: You copy the file test.fasta to 01_Rawdata with the following syntax

cp [source file] [target_directory]/

For example

cp test.fasta 01_Rawdata/

Copy file to another file, using the syntax

cp [source_file] [new_file_name]

For example

cp test.fasta test_2.fasta

You can copy a file to a new file in the directory by using the following syntax

```
cp [source_file] [target_directory]/[new_file_name]
```

For example

cp test.fasta 01_Rawdata/another_test.fasta

To copying directory, use additional flag as follow

cp -r [source_directory] [new_directory_name]

The flag -r stands for recursive, i.e. all files and subdirectories in this directory are copied repeatedly. For example, 01_Rawdata already contains test.fasta, which we copied before, and we want to duplicate this directory.

cp -r 01_Rawdata/ 01_Rawdata_new

Moving files and directories

To copy files and directories, the command mv can be used. mv stands for move and is used to move files and directories in Linux. For example, move the file test_2.fasta to the directory O1_Rawdata_new with the following syntax

```
mv [file_to_move] [target_directory]
```

```
m∨ test_2.fasta 01_Rawdata_new/
```

Specifically, to move files and directories, no flags are required as with cp. So if we want to move O1_Rawdata_new to a subdirectory of O1_Rawdata, this can be done as follows

```
mv [source_file_or_dir] [target_file_or_dir]
```

m∨ 01_Rawdata_new/ 01_Rawdata

Moving file within the directory up to the current directory

```
mv [source_dir]/[source_file] .
```

The dot (.) stands for the current directory, which means you want to move something to the current directory. For example, we want to move the file another_test.fasta, which is in the directory Ol_Rawdata, to the current directory by typing

mv 01_Rawdata/another_test.fasta .

Deleting files and directories

Removing files and directories can be done with the command rm. rm stands for remove and is used to delete files and directories in Linux. It's simple and straightforward with the following syntax.

rm [file_to_delete]

For example, you are deleting file another_test.fasta

rm another_test.fasta

To delete directories, use additional flags

rm -rf [directory_to_delete]

The flag -r means that it does something recursive, which means that it deletes all files and subdirectories of the directory you want to delete. The flag f can help us delete some protected files and directories that you should think twice before deleting.

For example you want to delete 03_adapter_trimming directory

rm -rf 03_adapter_trimming

Or delete subdirectory 01_Rawdata_new by

```
rm -rf 01_Rawdata/01_Rawdata_new
```

Don't worry~ the 01_Rawdata is still with us

```
jiratchaya@DESKTOP-P2DD13C:~/Cpa_RNASeq$ ls -1
total 20
drwxr-xr-x 2 jiratchaya jiratchaya 4096 Mar 1 22:33 01_Rawdata
drwxr-xr-x 2 jiratchaya jiratchaya 4096 Mar 1 21:59 02_QC
-rw-r--r-- 1 jiratchaya jiratchaya 38 Mar 1 22:00 another_test.fasta
-rw-r--r-- 1 jiratchaya jiratchaya 38 Mar 1 21:59 test.fasta
-rw-r--r-- 1 jiratchaya jiratchaya 38 Mar 1 22:00 test_2.fasta
```

□ **Danger zone:** Be sure to check the path of the location where you want to delete something with the command rm -rf, otherwise you'll unintentionally delete necessary files or directories.



Remove everything with sudo privilege. From meme-arsenal.

Downloading file from URL

There are numerous ways to download a file from a URL via the command line on Linux, and two of the best tools for this task are wget and curl. Both tools have their advantages and disadvantages, depending on the download task at hand. However, in this workshop we'll mainly focus on downloading with curl.

For example, we want to download the latest (draft) genome assembly report of *Cyanophora paradoxa* from the NCBI genome database via curl as follows.

curl -0 https://ftp.ncbi.nlm.nih.gov/genomes/all/GCA/004/431/415/GCA_004

Expected output

```
jiratchaya@DESKTOP-P2DD13C:~/Cpa_RNASeq$ curl -0
https://ftp.ncbi.nlm.nih.gov/genomes/all/GCA/004/431/415/GCA_004431415.1
t
 % Total
            % Received % Xferd Average Speed
                                                Time
                                                        Time
                                                                 Time
Current
                                Dload Upload
                                                Total
                                                        Spent
                                                                 Left
Speed
100 61357 100 61357
                             0 21604
                                           0 0:00:02 0:00:02 --:--:--
                       0
21604
jiratchaya@DESKTOP-P2DD13C:~/Cpa_RNASeq$ ls -1
total 80
drwxr-xr-x 2 jiratchaya jiratchaya 4096 Mar 1 22:33 01_Rawdata
drwxr-xr-x 2 jiratchaya jiratchaya 4096 Mar 1 21:59 02_QC
-rw-r--r-- 1 jiratchaya jiratchaya 61357 Mar 1 22:56
GCA_004431415.1_ASM443141v1_assembly_report.txt
-rw-r--r-- 1 jiratchaya jiratchaya
                                     38 Mar 1 22:00 another_test.fasta
-rw-r--r-- 1 jiratchaya jiratchaya
                                     38 Mar 1 21:59 test.fasta
-rw-r--r-- 1 jiratchaya jiratchaya
                                     38 Mar 1 22:00 test_2.fasta
•
                                                                     •
```

Tips: The alternative way to retrieve genome information from NCBI, you can just go to <u>NCBI Genome Data Hub</u> and specify species name to get information. NCBI provides several routes to download files including curl!

https://www.ncbi.nlm.nih.gov/data-hub,	/genome/GCA_004431415.1/		A* 16 G \$=
Datasets Taxonomy Gene	Gene Command-line tools Documenta	lion	BET
192 (mar)	enome assembly SM443141v1 reference		Additional genomes Browse all Cyanophora paradoxa genomes (1)
_			BioProject
Download 🕞 datasets	surl	Actions	PRJNA479824 Cyanophora paradoxa long-read genome
Submitted GenBank sequence	GCA_004431415.1	:	sequencing and assembly
Taxon	Cyanophora paradoxa		Publications
Isolate	CCMP329		Showing 2 of 2
WGS project	QPMI01		Mar Drugs 2020
Assembly type	haploid		Mining Natural Product Biosynthesis in
Submitter	Rutgers, The State University		Eukaryotic Algae
Date	Feb 5, 2019		
View the legacy Assembly pag	e		DNA Res 2019 Analysis of an improved Cyanophora paradoxa genome assembly
Assembly statistic	cs		DC Price, et al.
	GenBank		View in PubMed

A genome assembly of *C. paradoxa* in NCBI genome data hub (Accessed: 1 March 2023)

Inspecting file

We'll inspect the assembly report file

GCA_004431415.1_ASM443141v1_assembly_report.txt that we just downloaded from NCBI

• Count how many lines in that file

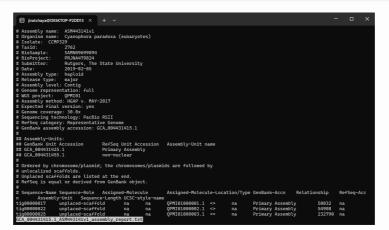
wc -l GCA_004431415.1_ASM443141v1_assembly_report.txt

```
jiratchaya@DESKTOP-P2DD13C:~/Cpa_RNASeq$ wc -1
GCA_004431415.1_ASM443141v1_assembly_report.txt
743 GCA_004431415.1_ASM443141v1_assembly_report.txt
```

• Print some contents at a time.

Now you will see the number of lines that fit on your screen, and you can scroll ${\sf up}$ and down with the arrow keys. Then press ${\sf q}$ when you have checked your file.

less GCA_004431415.1_ASM443141v1_assembly_report.txt



Example of inspecting a file with the less command. Users can scroll up and down with the arrow keys and exit by pressing q.

• Print top 10 lines of file

head GCA_004431415.1_ASM443141v1_assembly_report.txt



The first 10 lines of C. paradoxa assembly report file

• Print bottom 10 lines of file

tail GCA_004431415.1_ASM443141v1_assembly_report.txt

jiratchaya@DESK	TOP-P2DD13C:~/Cpa_RNASeq	s tail G	CA_004431415.1_A	5M443141v1_assemt	oly_repor	t.txt			
tig00022099	unplaced-scaffold		na QPMI0100	90711.1 <>		Primary	Assembly	30184	
tig00022104	unplaced-scaffold		na QPMI0100	90712.1 <>		Primary	Assembly	69684	
tig00001123	unlocalized-scaffold	MT	Mitochondrion	QPMI01000266.1		na -	non-nuclear	79456	
	unlocalized-scaffold	Pltd	Chloroplast	QPMI01000005.1		na	non-nuclear	1319169	
tig00021848	unlocalized-scaffold	Pltd	Chloroplast	QPMI01000704.1		na	non-nuclear	8078	
tig00021860	unlocalized-scaffold	Pltd	Chloroplast	QPMI01000702.1		na	non-nuclear	2191	
tig00021925	unlocalized-scaffold	Pltd	Chloroplast	QPMI01000705.1		na	non-nuclear	48847	
tig00022054	unlocalized-scaffold	Pltd	Chloroplast	QPMI01000706.1		na	non-nuclear	78756	
tig00022061	unlocalized-scaffold	Pltd	Chloroplast	QPMI01000708.1		na	non-nuclear	45352	
tig08822078	unlocalized-scaffold	Pltd	Chloroplast	OPMT01000707.1		na	non-nuclear	6648	na

The last 10 lines of C. paradoxa assembly report file.

• Print only lines with a specific pattern of word.

For example, we'll print only the lines contain the word "Chloroplast"

grep "Chloroplast" GCA_004431415.1_ASM443141v1_assembly_report.txt



Extracted lines with a specific word "Chloroplast" in assembly report file.

Show latest commands we used

You can simply press arrow keys up or down to see your latest commands that you typed in the terminal.

Another way to see the latest command by typing below in the terminal

history

History is able to keep track of the command lines you use, associate any data with each line, and use information from previous lines when writing new lines.

Shortcut: Tab Completion

When typing file or directory names, it's easy to mistype. Instead, we can use 'tab' to complete what we want to type. The shell will try to fill in the rest of a directory or file name if you press tab after typing.

Have no idea what this command can do

Basically, the built-in Linux system commands store usage in their command manual. You can call man followed by a command you want to learn more about. for example man curl.



A manual page of curl. Users can scroll using arrow keys up and down, and quit reading by press q.

Maintaining Long-Running Jobs with tmux

When we run programs through the Unix shell, they run until they terminate successfully or are terminated with an error. Multiple processes running simultaneously on your computer, such as system files, web browser, email application, bioinformatics programs, and so on. In bioinformatics, we often work with processes that run for a long period of time. Therefore, it's important that we know how to work with processes and manage them using the Unix shell. In this section, we'll learn the basics of dealing with processes.

In addition, processes are also terminated if the connection to the servers is interrupted, the network connection drops immediately, or the power fails. Since we're constantly working with remote computers in our daily work in bioinformatics, we need a way to prevent the accidental termination of longrunning applications. Leaving the local terminal's connection to a remote computer open while a program is running is an unsafe solution, even the most reliable networks can experience short outages.



How tmux increase you pruductivity :/ (From <u>Billy uses tmux</u> in Reddit)

Some software offers the user the possibility to run their work as a background process, e.g. Nohup, Screen and Tmux. In this workshop, we propose **Terminal Multiplexer (Tmux)**, which allows you to create a session with multiple windows, each of which can run its own processes. The Tmux sessions are persistent, which means that all the windows and their processes can be easily restored by reattaching the session.

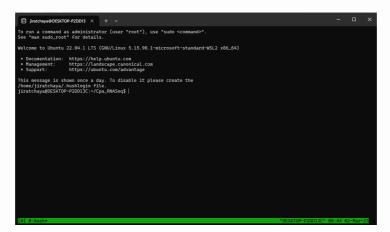
When Tmux is running on a remote machine, you can maintain a persistent session that isn't lost when the connection drops or you close your terminal window to go home (or even exit your terminal programme). Rather, all Tmux sessions can be reattached to the terminal you're currently at - simply log back into the remote machine via SSH and reattach the Tmux session. All windows remain undisturbed and all processes continue to run.

A simple usage of Tmux

Open a terminal and use the following command

tmux

You see a command prompt as usual, but you now see a taskbar-style menu at the bottom of the terminal that contains something like bash o *. The asterisk indicates that this is your active window.



Tmux windows

Detach a session

This allows you to leave the tmux session, but it continues to run in the background. Just press the key

[ctr] + b] + d

Your terminal will print

```
jiratchaya@DESKTOP-P2DD13C:~/Cpa_RNASeq$ tmux
[detached (from session 0)]
```

This should take you back to a standard prompt. Remember that the Tmux session continues in the background, and you can recall it at any time.

Name the Tmux session

You may find it helpful to name your sessions with meaningful titles to keep things organized. Let's try naming your first session with Tmux.

You can name it anything that we want, but in this case I will name it 'process2'. Enter the following command:

tmux new -s process2

You should now have a new Tmux session running. If you look in the lower left area of the window, you will see the name of your session rather than the generic 'bash'.

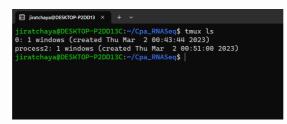
List tmux sessions

What happened to your session? It is still running in the background. You can reopen the session by name or number ID, but what if you forgot the session name?

There is a list function built into tmux:

tmux ls

This lists all your current tmux sessions. When you run it, you get output like this:



List of running tmux sessions.

Reenter (aka reattach) a session in Tmux

To reopen your tmux session, you can use the tmux command with the attach or attach-session option as follows:

tmux a -t [session_name]

For example, we'll reenter to the process2 session.

tmux a <mark>-t</mark> process2

Exit tmux when finish running

Quitting tmux is exactly the same as quitting the standard terminal by pressing the keys Ctrl+d or by entering

exit

Resources

- Buffalo, V. (2015). *Bioinformatics data skills: Reproducible and robust research with open source tools.* "O'Reilly Media, Inc.".
- <u>Introduction to the command line interface</u> by Harvard Chan Bioinformatics Core (Accessed on 27 Feb 2023).
- <u>Introducing the Shell</u>, from the course Introduction to the Command Line for Genomics in bioinformatics-core-shared-training (Accessed on 28 Feb 2023)
- <u>Bash cheat sheet</u> from RehanSaeed GitHub repository (Accessed on 1 March 2023).
- <u>Getting Started with Tmux [Beginner's Guide]</u>. By linuxhandbook.com (Accessed on 2 March 2023)

Data Retrieval with NCBI SRA Toolkit

NCBI (Natianal Center for Biotechnology Information) is a major source of biological databases related to life and health sciences research, as well as a major source of bioinformatics tools and services. NCBI hosts various types of biological data submitted by researchers from around the world, such as <u>GenBank</u> for nucleotide sequence submissions, <u>Sequence Read Archive (SRA)</u> for raw sequence data, <u>Genome</u> for submitting full or draft genomes, <u>Gene Expression</u> <u>Omnibus (GEO)</u> for quantitative gene expression data sets, and many more.

<u>NCBI SRA toolkit</u> is a set of utilities for downloading, viewing, and searching large amounts of high-throughput sequencing data from the NCBI SRA database.

SRA toolkit can

- Effectively download the large volume of high-throughput sequencing data
- Convert SRA file into other biological file format
- Retrieve a small subset of large files
- Search within SRA files and fetch specific sequences



Screenshots of NCBI Sequence Read Archives

What is Sequence Read Archives (SRA)

The Sequence Read Archive (SRA) is the largest publicly accessible repository for high-throughput sequencing data. SRA accepts data from all areas of sequencing projects as well as metagenomics and environmental studies. Sequencing data may be isolated from a single species or from multiple species as in metagnomics studies.

SRA also refers in the file description to the format defined by NCBI for NGS data in the SRA database. All data submitted to NCBI must be stored in SRA format and can be converted back to a FASTQ, FASTA, or BAM file depending on the original submission by the researchers. Here, the SRA Toolkit provides tools for downloading data, converting various data formats to SRA format and vice versa, and extracting SRA data to other formats.

Researchers often use SRA data to make discoveries and conduct reproducible research. Data sets can be compared using the SRA web interface. However, if you want to download files for local use on your computer, you should use a command line interface, and the SRA Toolkit is highly recommended by NCBI.

Searching RNA-Sequencing datasets in NCBI

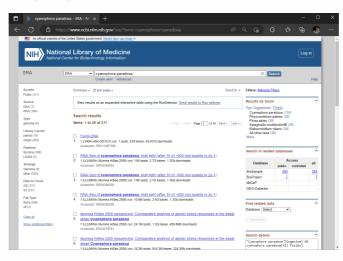
The databases in NCBI are linked by some common features. This means that you can start wherever you have your research problems in NCBI. In this workshop, we will investigate transcriptional changes during light exposure of the alga *Cyanophora paradoxa*, a representative species of Glaucophytes. For more information about this alga, please see this article in Science.

In this workshop we'll retrieve transcriptome sequencing libraries of C. paradoxa under the normal light and dark conditions. This dataset is generated by (Knopp et al. 2020).

Activity

You can easily search the SRA database for any keywords of interest related to your research. In this context, we search for all SRA studies related to *C. paradoxa* and see what SRA provides us. Note that the SRA database contains not only transcriptome studies, but also genomes and metagenomes.

1. Go to SRA database: <u>https://www.ncbi.nlm.nih.gov/sra</u>, and search for 'cyanophora paradoxa'.



Screenshot of the search result of *C. paradoxa* in the database NCBI SRA. Here you can see all sequencing libraries of this species that have been submitted to NCBI. You can specify which items are of interest or customize the search using the filter box on the left and right sides of the screen.

 We'll adjust our selection using the tool in the SRA database, the SRA Run Selector, as follows.

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Browsing sequencing data in NCBI SRA database

3. In the SRA Run Selector, you can customize the filters based on the metadata columns of all runs. In this case, we filter the SRA runs based on the assay type as RNA-Seq and select only paired-end sequencing data as follows.

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Customizing filters in SRA Run Selector

4. Then you can select which runs you want to download and perform analysis. In this workshop we'll select *C. paradoxa* RNA-Seq reads from SRR8306028, SRR8306029, SRR8306032, SRR8306033, SRR8306034 and SRR8306035.

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		0.5	SRR11829735	PRJNA634446	SAMN14996509	RNA-Seq	243	2.65 G	1.25 Gb	HEINRICH HEINE UNIVERSITY	SRX8380258
		0.0	SRR363339		SAMIN00750200	RNA-Seq	245	15.33 G	8.98 Gb	RUTGERS UNIVERSITY	SRX104482
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		8	SRR8306029	PRJNA509798	SAMIN10588740	RNA-Seq	300	3.36 G	1.75 Gb	HHU DUESSELDORF	SRX5120514
		9	SRR8306032	PRJNA509798	SAMN10588739	RNA-Seq	300	3.38 G	1.77 Gb	HHU DUESSELDORF	SRX5120511
		10	SRR8306033	PRJNA509798	SAMIN10588739	RNA-Seq	300	3.39 G	1.76 Gb	HHUDUESSELDORF	SRX5120510
		11	SRR8306034	PRJNA509798	SAMN10588740	RNA-Seq	300	3.44 G	1.77 Gb	HHU DUESSELDORF	SRX5120509
		12	SRR8306035	PRJNA509798	SAMN10588739	RNA-Sep	300	3.44 G	1.78 Gb	HHU DUESSELDORF	SRX5120508

Exporting the selected metadata in SRA Run Selector.

5. The downloaded metadata is in comma-separated file format. So you can open them with spreadsheet programs like Microsoft Excel on your local laptop. The metadata looks like this.

Downloading SRA runs using fasterq-dump

fasterq-dump are tools in the SRA toolkit used to connect from our remote server to the NCBI server and download sequencing data from SRA.

According to the <u>NCBI sra-tools' guideline</u>, using fasterq-dump in combination with another tool, prefetch, is the better way to download data because prefetch can be invoked at any time if the previous download accidentally failed. So it is not necessary to start the download from the beginning.

However, prefetch can sometimes be skipped if you want to download a small amount of data. In this workshop we'll use only fasterq-dump to download and process SRA file format to FASTQ file for further analysis.

Activity

We'll do the following command in Terminal or MobaXterm, by access to the username and password that we've provided.

- For mobaXterm, enter to your session.
- For terminal, type

```
ssh <username>@<server IP address>
```

Now we download RNA-Seq libraries from *C. paradoxa* using the SRA accessions listed in the first column of the metadata above using the following command.

Activate analysis environment

conda activate ncbi

Go to working directory

cd ~/Cpa_RNASeq/01_Rawdata

Run fasterq-dump

```
fasterq-dump --threads 2 --progress \
SRR8306028 SRR8306029 SRR8306032 \
SRR8306033 SRR8306034 SRR8306035
```

From the fasterq-dump command,

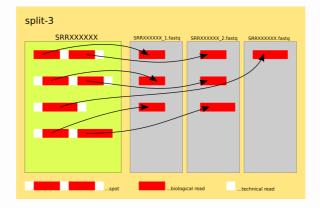
--threads refer to how many threads to use (default = 6).

--progress force the terminal to print the progress of downloading and processing file to the screen.

Expected output files.

01_F	Rawdata
\vdash	SRR8306028_1.fastq
\vdash	SRR8306028_2.fastq
\vdash	SRR8306029_1.fastq
\vdash	SRR8306029_2.fastq
\vdash	SRR8306032_1.fastq
\vdash	SRR8306032_2.fastq
\vdash	SRR8306033_1.fastq
\vdash	SRR8306033_2.fastq
\vdash	SRR8306034_1.fastq
\vdash	SRR8306034_2.fastq
\vdash	SRR8306035_1.fastq
L	SRR8306035_2.fastq

By default, fasterq-dump processes a single SRA file format of paired-end reads by splitting reads into forward (*_1.fastq) and reverse (*_2.fastq), if singletons (unpaired between forward and reverse reads) present, it will be written to another fastq file as described in this figure.



Sequence read processing by fasterq-dump using default parameters. Figure adopted from <u>https://github.com/ncbi/sra-tools/wiki/HowTo:-fasterq-dump</u>.

Reference Sources

- Price, Dana C., et al. "*Cyanophora paradoxa* genome elucidates origin of photosynthesis in algae and plants." *Science* 335.6070 (2012): 843-847. https://doi.org/10.1126/science.1213561.
- <u>SRA Toolkit: the SRA database at your fingertips</u> from NCBI Insights. Accessed 4-Mar-2023.
- <u>How to use NCBI SRA Toolkit effectively</u> by Renesh Bedre, Data science blog. Accessed 4-Mar-2023.
- HowTo: fasterq dump by NCBI sra-tools GitHub Wiki. Accessed 4-Mar-2023.

RNA-Seq Data Quality Control

What is FASTQ file format

Next-generation sequencing and data analysis projects typically begin with the processing of sequence read data and their quality tags from the sequencer in FASTQ format. The FASTQ format is the most commonly used format in sequence analysis and is generated by a sequencer. The FASTQ file contains the sequence data from the clusters that pass the filter of a flow cell. Many analysis tools require this format because it contains much more information than FASTA. In this workshop, we will mainly explain the FASTQ file format, which comes from the Illumina sequencer.

The FASTQ format is similar to the fasta format, but differs in syntax and in the integration of quality values. Each sequence requires at least 4 lines:

Example of FASTQ file format

- 1. A sequence identifier with information about the sequencing run and the cluster. The exact contents of this line vary by based on the BCL to FASTQ conversion software used.
- 2. The sequence (the base calls; A, C, T, G and N).
- 3. A separator, which is simply a plus (+) sign.
- 4. The base call quality scores. These are Phred +33 encoded, using ASCII characters to represent the numerical quality scores.

The FastQ sequence descriptor generally follows a specific format that includes all information about the sequencer and its position on the flow cell. The sequence descriptor also follows a specific format and contains information about the sample information.

FASTQ sequence descriptor, particulary in Illumina sequence reads look like:

@HWUSI-EAS100R:6:73:941:1973#0/1

where

HWUSI- EAS100R	The unique instrument name
6	Flowcell lane
73	Tile number within the flowcell lane
941	'x'-coordinate of the cluster within the tile
1973	'y'-coordinate of the cluster within the tile
# 0	Index number for a multiplexed sample (0 for no indexing)
/1	The member of a pair, /1 or /2 (paired-end or mate-pair reads only)

As mentioned earlier, line 4 contains the quality score of the nucleotide at the same position. The quality scores are represented by the code ASCII, which indicates how confident of the correctly called base is.

We can calculate the quality score of a base, if P is the error probability, then:

$Q = -10 log_{10}(P)$

The following figure shows the representative ASCII code for the score value. Base quality scoring for analysis is important when identifying types of genomic variation such as SNPs, but it is also an indicator of the overall quality of the sequencing as well.

2	P error	ASCII	Q	P error	ASCII	Q	P error	ASCII	Q	P error	ASCII
5	1.00000	33 !	11	0.07943	44 ,	22	0.00631	55 7	33	0.00050	66 B
1	0.79433	34 "	12	0.06310	45 -	23	0.00501	56 8	34	0.00040	67 C
2	0.63096	35 #	13	0.05012	46 .	24	0.00398	57 9	35	0.00032	68 D
3	0.50119	36 \$	14	0.03981	47 /	25	0.00316	58 :	36	0.00025	69 E
4	0.39811	37 %	15	0.03162	48 0	26	0.00251	59 ;	37	0.00020	70 F
5	0.31623	38 &	16	0.02512	49 1	27	0.00200	60 <	38	0.00016	71 G
5	0.25119	39 '	17	0.01995	50 2	28	0.00158	61 =	39	0.00013	72 H
1	0.19953	40 (18	0.01585	51 3	29	0.00126	62 >	40	0.00010	73 I
	0.15849	41)	19	0.01259	52 4	30	0.00100	63 ?	41	0.00008	74 J
	0.12589	42 *	20	0.01000	53 5	31	0.00079	64 @	42	0.00006	75 K
	0.10000	43 +	21 Lumina	0.00794	54 6	32	0.00063	65 A			
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C	II_BASE=6 P_error 1.00000	4 Old Ill ASCII 64 0	lumina Q 11	P_error 0.07943	ASCII 75 K	Q 22	P_error 0.00631	ASCII 86 V	33	0.00050	97 a
5C 2	II_BASE=6 P_error	4 Old Ill ASCII	Lumina Q 11 12	P_error	ASCII	Q	P_error	ASCII	100 CT	i i e 🗖 e stadio e e e	97 a 98 b
SC 2) 1 2	II_BASE=6 P_error 1.00000 0.79433	64 01d I11 ASCII 64 0 65 A	lumina Q 11	P_error 0.07943 0.06310	ASCII 75 K 76 L	Q 22 23	P_error 0.00631 0.00501	ASCII 86 V 87 W	33 34	0.00050	97 a 98 b 99 c
SC 2 0 1 2 3	II_BASE=6 P_error 1.00000 0.79433 0.63096	64 Old Ill ASCII 64 @ 65 A 66 B	2 11 12 13	P_error 0.07943 0.06310 0.05012	ASCII 75 K 76 L 77 M	Q 22 23 24	P_error 0.00631 0.00501 0.00398	ASCII 86 V 87 W 88 X	33 34 35	0.00050 0.00040 0.00032	97 a 98 b 99 c 100 d
SC 2	II_BASE=6 P_error 1.00000 0.79433 0.63096 0.50119	64 Old Ill ASCII 64 0 65 A 66 B 67 C	11 12 13 14	P_error 0.07943 0.06310 0.05012 0.03981	ASCII 75 K 76 L 77 M 78 N	Q 22 23 24 25	P_error 0.00631 0.00501 0.00398 0.00316	ASCII 86 V 87 W 88 X 89 Y	33 34 35 36	0.00050 0.00040 0.00032 0.00025	97 a 98 b 99 c 100 d 101 e
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	II_BASE=6 P_error 1.00000 0.79433 0.63096 0.50119 0.39811 0.31623	64 Old Ill ASCII 64 @ 65 A 66 B 67 C 68 D 69 E	Lumina Q 11 12 13 14 15 16	P_error 0.07943 0.06310 0.05012 0.03981 0.03162 0.02512	ASCII 75 K 76 L 77 M 78 N 79 O 80 P	Q 22 23 24 25 26 27	P_error 0.00631 0.00501 0.00398 0.00316 0.00251 0.00200	ASCII 86 V 87 W 88 X 89 Y 90 Z 91 [33 34 35 36 37 38	0.00050 0.00040 0.00032 0.00025 0.00020 0.00016	97 a 98 b 99 c 100 d 101 e 102 f 103 g
	II_BASE=6 P_error 1.00000 0.79433 0.63096 0.50119 0.39811 0.31623 0.25119	4 Old III ASCII 64 @ 65 A 66 B 67 C 69 E 70 F	2 11 12 13 14 15 16 17	P_error 0.07943 0.06310 0.05012 0.03981 0.03162 0.02512 0.01995	ASCII 75 K 76 L 77 M 78 N 79 O 80 P 81 Q	Q 22 23 24 25 26 27 28	P_error 0.00631 0.00501 0.00398 0.00316 0.00251 0.00220 0.00158	ASCII 86 V 87 W 88 X 89 Y 90 Z 91 [92 \	33 34 35 36 37 38 39	0.00050 0.00040 0.00032 0.00025 0.00020 0.00016 0.00013	ASCII 97 a 98 b 99 c 100 d 101 e 102 f 103 g 104 h 105 i
5 2 2 3 1 5 5 7 3 3	II_BASE=6 P_error 1.00000 0.79433 0.63096 0.50119 0.39811 0.31623 0.25119 0.19953	64 Old III ASCII 64 @ 65 A 66 B 67 C 68 D 69 E 70 F 71 G	umina Q 11 12 13 14 15 16 17 18	P_error 0.07943 0.06310 0.05012 0.03981 0.03162 0.02512 0.01585	ASCII 75 K 76 L 77 M 78 N 79 O 80 P 81 Q 82 R	Q 22 23 24 25 26 27 28 29	P_error 0.00631 0.00501 0.00398 0.00316 0.00251 0.00200 0.00158 0.00126	ASCII 86 V 87 W 88 X 89 Y 90 Z 91 [92 \ 93]	33 34 35 36 37 38 39 40	0.00050 0.00040 0.00032 0.00025 0.00020 0.00016 0.00013 0.00010	97 a 98 b 99 c 100 d 101 e 102 f 103 g 104 h

Tables converting between integer Q scores, ASCII characters and error probabilities. Figure adopted from

https://www.drive5.com/usearch/manual/quality_score.html

What software use FASTQ

To date, Almost NGS analysis software requires FASTQ format. For example:

- QC such as <u>fastQC</u> used FASTQ to determine how good of the sequence read library, generate an informative report, and also determining the presence of adapter sequences which can also be trimmed by some integrated QC tools such as <u>FASTP</u>.
- Aligners such as <u>bowtie2</u>, <u>BWA</u>, <u>STAR</u>, and so on, use reads, and quality sometimes, to align to the reference sequence. The mapping information can be further used for quantifying expression, constructing sequence assembly, and variant calling.
- De novo assembly tools, for example <u>Trinity</u>, <u>Spades</u>, <u>Velvet</u>, etc., also use FASTQ to construct contig library and scaffolding. Some de novo assembler tools not only use FASTQ to contruct draft assembly but also used in the polishing process to refine assembly, such as <u>Flye</u>, <u>Unicycler</u>, <u>Canu</u>, etc.

Quality assessment using FastQC

FastQC is designed for quality control of raw sequence data from highthroughput sequencing technology. It provides a modular set of analyses that you can use to get a quick overview of the quantity and quality of your data, and to help you decide on the raw data whether you should perform adapter or lowquality read trimming or whether you can perform further analyses. For sequence reads that require adapter trimming before further analysis, we recommended to assessing the quality both before and after trimming.

Most sequencers will generate a QC report as part of their analysis pipeline, but this is usually only focused on identifying problems which were generated by the sequencer itself. FastQC aims to provide a QC report which can spot problems which originate either in the sequencer or in the starting library material.

Activity

The following will perform on you user account by activating your working environment at first.

conda activate qc

Then, create a directory for QC result before adapter trimming

mkdir 02_QC/fastQC_before_trim

Run FastQC all file at once. Here, we'll use a wildcard *.fastq to select all FASTQ files in 01_Rawdata directory. We also specify number of CPU threads in --threads and QC output files in 02_QC/fastQC_before_trim using --outdir argument.

```
fastqc --outdir 02_QC/fastQC_before_trim \
--threads 2 \
/opt/Cpa_RNASeq/01_Rawdata/*.fastq
```

Estimated time: ~10min

Interpreting FastQC results

FastQC also provided excellent explanation of each analysis step in their documentation. So we encouraged you to learn more at their <u>web page</u> along with the <u>documentation</u>.

The analysis in FastQC is performed by a series of analysis modules. The left hand side of the main interactive display or the top of the HTML report show a summary of the modules which were run, and a quick evaluation of whether the results of the module seem entirely normal (green tick), slightly abnormal (orange triangle) or very unusual (red cross).



FastQC sidebar

Basic statistics

The Basic Statistics module generates some simple composition statistics for the file analysed.

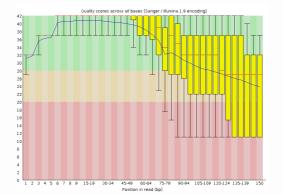
Measure	Value
Filename	SRR8306028_1.fastq
File type	Conventional base calls
Encoding	Sanger / Illumina 1.9
Total Sequences	11244423
Total Bases	1.6 Gbp
Sequences flagged as poor quality	0
Sequence length	150
%GC	62

Basic statistics of SRR8306028.

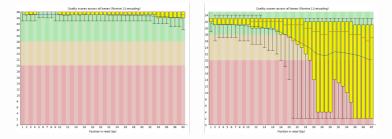
- Filename: The original filename of the file which was analysed
- File type: Says whether the file appeared to contain actual base calls or colorspace data which had to be converted to base calls
- **Encoding:** Says which ASCII encoding of quality values was found in this file.
- Total Sequences: A count of the total number of sequences processed. There are two values reported, actual and estimated. At the moment these will always be the same. In the future it may be possible to analyse just a subset of sequences and estimate the total number, to speed up the analysis, but since we have found that problematic sequences are not evenly distributed through a file we have disabled this for now.
- **Sequence Length:** Provides the length of the shortest and longest sequence in the set. If all sequences are the same length only one value is reported. %GC: The overall %GC of all bases in all sequences

Per Base Sequence Quality

The Per base sequence quality plot shows an overview of the range of quality values across all bases at each position in the FastQ file.



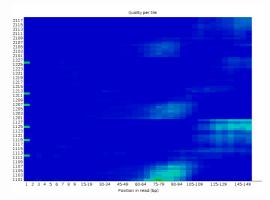
Per Base Sequence Quality plot of SRR8306028. In which the central red line is the median value. The yellow box represents the interquartile range (25-75%). The upper and lower whiskers represent the 10% and 90% points. The blue line represents the mean quality. The higher the score, the better the base call, i.e., the box plots fall into the very good quality area (green background), the mediocre quality area (orange background), and the poor quality area (red background). The following figures show a comparison of the good and poor quality results of Illumina sequencing technology.



A comparison of good (left) and bad (right) per base sequence quality plots. Figures adopted from example reports in https://www.bioinformatics.babraham.ac.uk/projects/fastqc.

Per tile sequence quality

This plot is specific to Illumina sequencing libraries and shows colour shading of quality score by position on the flow cell. The colours are on a scale from cold to hot, with cold colours representing positions where the quality was at or above average for that base in the run, and hotter colours indicating that a tile had worse qualities than other tiles for that base. In the example below, you can see that certain tiles have consistently poor quality. A good chart should be blue throughout.



Per tile sequence quality plot of SRR8306028.

Per Sequence Quality Scores

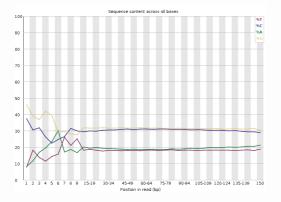
The per sequence quality score report allows you to see if a subset of your sequences have universally low quality values. It is often the case that a subset of sequences will have universally poor quality, often because they are poorly imaged (on the edge of the field of view etc), however these should represent only a small percentage of the total sequences.



Per Sequence Quality Scores plot of SRR8306028.

Per Base Sequence Content

Per Base Sequence Content plots out the proportion of each base position in a file for which each of the four normal DNA bases has been called. The plot shows the quality of nucleotide A T C and G separately into 4 lines.



Per Base Sequence Content plot of SRR8306028.

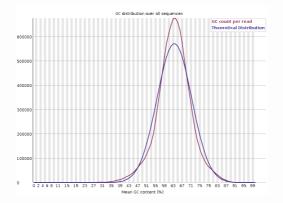
Usually ambiguous base values are found at the beginning of the read. Libraries made with random hexamer primers, with almost all RNA-Seq libraries using them, and those that were fragmented libraries. This bias does not affect an absolute sequence, but provides enrichment of a number of different K-mers at the 5' end of the reads. While this is a true technical bias, it cannot be corrected by trimming and does not appear to affect downstream analysis in most cases. However, a warning or error is generated in this module. This module issues a warning if the difference between A and T, or G and C is greater than 10% in any position.

Per Sequence GC Content

This module measures the GC content over the entire length of each sequence in a file and compares it to a normal distribution of GC content. Normally, one would expect an approximately normal distribution of GC content, where the central peak corresponds to the total GC content of the underlying genome of interest.

The skewness of the distribution may indicate some unusual events such as

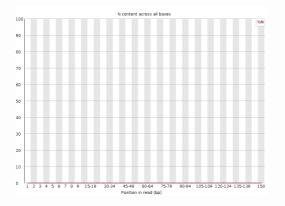
contamination or systematic bias in your sequencing library. However, the GC content signature of different organisms may depend on their nature.



Per Sequence GC Content plot of SRR8306028.

Per base N content

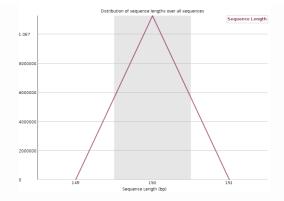
This module represents the percentage of base calls at each position for which an N was called. The 'N' base is found when the sequencer is not able to make a confident base call, then it will normally substitute an N.



Per base N content plot of SRR8306028.

Sequence Length Distribution

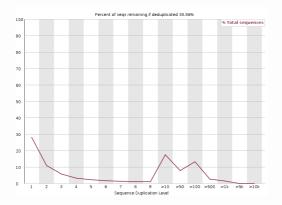
This module generates a histogram of distribution of sequence reads in the file which was analyzed.



Sequence Length Distribution plot of SRR8306028.

Sequence Duplication Levels

This module counts the degree of duplication for every sequence in a library and creates a plot showing the relative number of sequences with different degrees of duplication. A low level of duplication may indicate a very high level of coverage of the target sequence, but a high level of duplication is more likely to indicate some kind of enrichment bias (eg PCR over amplification).



Sequence Duplication Levels plot of SRR8306028.

Overrepresented sequences

This module lists all sequences that make up more than 0.1% of the first 100,000 sequences examined. For each overrepresented sequence, the program searches for matches in a database of common impurities and reports the best match found. However, finding a hit doesn't mean that this is the source of the contamination, but may point you in the right direction.

Adapter Content

This plot shows the cumulative percentage of adapter sequences used for sequencing this library at each position. Most adapter sequences found in Illumina RNA-Seq libraries are Illumina Universal Adapters. This module issues a warning if a sequence is present in more than 5% of all reads. This module issues a warning if a sequence is present in more than 10% of all reads. If the adapter sequence is present in more than 1% of the sequence library, adapter

trimming is considered.

.00	% Adapter	Illumina Universal Adapter
30		flumina Small RNA 3' Adapter flumina Small RNA 5' Adapter Nextera Transposase Sequence
30		PolyA PolyG
70		
50		
50		
10		
30		
10		
1 2 3 4 5 6 7 8 9 12 13 18 19 24 25 30	31 36:37 42:43 48:49 54:55 60:61 66:67 72:73 78:79 84:6	35 90-91 96-97 104-105 114-115 124-125 134-135

Adapter Content of SRR8306028.

Activity

To further combine all the QC results into a single interactive HTML file, we'd suggested to use multiqc software to combine it.

conda activate qc

Then, run multiqc

```
multiqc --filename QCreport_before_trim \
--outdir 02_QC/ \
--dirs 02_QC/fastQC_before_trim/
```

Estimated time: < 1 min

Output files:

```
- QCreport_before_trim_data
```

```
├── multiqc_citations.txt
```

```
├── multiqc_data.json
```

```
├─ multiqc_fastqc.txt
```

```
├── multiqc_general_stats.txt
```

```
├─ multiqc.log
```

```
└── multiqc_sources.txt
```

```
- QCreport_before_trim.html
```

Adapter Trimming with Cutadapt

Cutadapt is a tool to remove sequencing adapters, primers, poly-A tails and other types of unwanted sequence from your high-throughput sequencing reads. Cutadapt supports both FASTQ and FASTA file format for trimming.

Several types of sequencing adapters have been used nowaday. We have to know which adapter found in our sequencing library. Fortunately, Illumina provide a manual of <u>Illumina Adapter Sequences</u> that used in different types of sequencing. As mentioned, most of RNA-Seq library sequenced by Illumina used Illumina TruSeq Single Indexes, which is AGATCGGAAGAGCACACGTCTGAACTCCAGTCA and AGATCGGAAGAGCGTCGTGTAGGGAAAGAGTGT flanked at the 5' end of forward reads and 3' end of reverse reads, respectively.

Fortunately, the dataset that will be used is free of apadter sequences examined from the <u>adapter content</u> from FastQC result. *So this command will just*

show as a demo for your future project.

An example command of Cutadapt as follow.

```
cutadapt --cores 2 \
-u 10 -U 10 \
-a AGATCGGAAGAGCACACGTCTGAACTCCAGTCA \
-A AGATCGGAAGAGCGTCGTGTAGGGAAAGAGTGT \
-o <output_forward.fastq> \
-p <output_reverse.fastq> \
<input_forward.fastq> <input_reverse.fastq> \
```

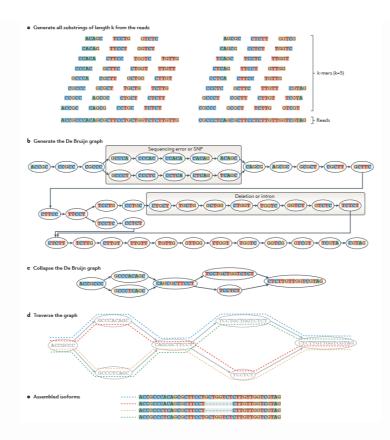
According to the command, we specify number of CPU threads in --cores. We remove 10 bases directly from each end of read, -u for forward and -u for reverse reads. Then we specify the adapter sequences as mentioned above in a and A. And the paths for forward and reverse reads output files in -o and -p, respectively.

Reference Sources

- <u>Quality Control of FASTQ files</u> from Harvard Chan Bioinformatics Core (HBC) training (Accessed on 1 Mar 2023).
- <u>FastQC official website</u> from Babraham Bioinformatics (Accessed on 1 Mar 2023).
- <u>FastQC Documentation</u> from Babraham Bioinformatics (Accessed on 1 Mar 2023).
- Cutadapt 4.2 Documentation (Accessed on 1 Mar 2023).
- Illumina Adapter Sequences (Accessed on 1 Mar 2023).

De novo Assembly with Trinity

Trinity is a promising tool for de novo full-length transcriptome assembly that continually developed since 2011. Trinity assembles reads by constructs many individual de Bruijn graphs, each representing the transcriptional complexity at a given gene or locus, that originated from the different nucleotide in the same position, and then processes each graph independently to extract full-length splicing isoforms and to tease apart transcripts derived from paralogous genes (Grabherr et al. 2011; Haas et al. 2013). Each assembled contig is will refer to a transcript.



Overview of the concept of de novo transcriptome assembly. Clean reads are divided into k-mers, i.e., in this figure k = 5, which means that a read is divided into many fragments, each fragment containing 5 bases. Then, de Bruijn graphs are stitched from a pool of billions of kmers (b). During sequencing, read fragments originating from the same spot or derived from the same gene may even have a nucleotide change at the same position, which may be either true polymorphisms or sequencing errors, so that similar k-mer sequences are joined together and routing to adjacent k-mers (c). The bulges in the graphs represent variations within the graph complex. Each graph complex represents a gene that can split into many transcript isoforms during traversion (d) to eventually obtain the assembled transcript library (e). For more details, please see Martin and Wang (2011).

Trinity can construct genomes without genome information and enables transcript construction in non-model organisms where genome assembly is not yet available, or that do not achieve successful chromosome-level or full assembly. Downstream processes, such as transcript assembly completeness analysis, transcript abundance estimation, and identification of differentially expressed genes, can also be performed with Trinity and its built-in utilities commands.

For de novo assembly in fast and efficient way for limited computational resources available, we prepared the downsized reads that derived from the SRA accessions that we already retrieved from NCBI SRA database from the previous chapter.

During de novo assembly, the longest and the heaviest computation resource

required for constructing and stitching billions of de Bruijn graphs. Very deep sequencing libraries may failed of these processes. Therefore, normalizing or downsizing sequence reads before de novo assembly is efficient way to proceed it. In brief, we downsized sequence reads using built-in Trinity command insilico_read_normalization.pl as follow. *This just inform all participant to the source of raw data that they will perform assembly*.

```
insilico_read_normalization.pl \
--seqType fq \
--JM 100G \
--max_cov 10 \
--left 01_Rawdata/*_1.fastq \
--right 01_Rawdata/*_2.fastq \
--pairs_together \
--CPU 50 \
--output 03_assembly
```

Trinity insilico_read_normalization.pl uses forward and reverse reads input from --left and --right parameters by reduce the maximum coverage depth (--max_cov) observed to 10x, and retain only paired reads in --pairs_together.

The output file given as follow:

```
-rw-r--r- 1 jiratchaya jiratchaya 1005390788 Mar 8 10:34 left.fq
-rw-r--r-- 1 jiratchaya jiratchaya 1005390788 Mar 8 10:34 right.fq
```

Running Trinity

Trinity is run via the script Trinity:

```
Trinity --seqType fq \
--max_memory 6G \
--CPU 2 \
--left /opt/Cpa_RNASeq/insilico_norm_reads/left.fq \
--right /opt/Cpa_RNASeq/insilico_norm_reads/right.fq \
--output Trinity_2023-03-08
```

Estimated time usage: ~24 hr

By this command, Trinity take the input forward and reverse reads from --left and --right, respectively. By default, de novo assembly with Trinity will perform in silico read normalization by itself. Since we have already normalized the sequence reads file prior to assembly, this step will skip the normalization step by add --no_normalize_reads to the command. All other arguments will use with the default parameters.

The given results in ~/Cpa_RNASeq/03_assembly directory are:

-rw-r--r-- 1 jiratchaya jiratchaya 77M Mar 8 11:34 Trinity_2023-03-08.Trinity.fasta -rw-r--r-- 1 jiratchaya jiratchaya 3.0M Mar 8 11:34 Trinity_2023-03-08.Trinity.fasta.gene_trans_map

• *.Trinity.fasta is the assembled transcript files

• *.Trinity.fasta.gene_trans_map is tab-separated file of trinit genes (left column) and the belonging transcript (right column)

```
jiratchaya@pslab1:~/Cpa_RNASeq/03_assembly$ head Trinity_2023-03-
08.Trinity.fasta.gene_trans_map
TRINITY_DN43810_c0_g1 TRINITY_DN43810_c0_g1_i1
```

```
      TRINITY_DN43890_c0_g1
      TRINITY_DN43890_c0_g1_i1

      TRINITY_DN43840_c0_g1
      TRINITY_DN43840_c0_g1_i1

      TRINITY_DN43847_c0_g1
      TRINITY_DN43847_c0_g1_i1

      TRINITY_DN43815_c0_g1
      TRINITY_DN43815_c0_g1_i1

      TRINITY_DN43872_c0_g1
      TRINITY_DN43872_c0_g1_i1

      TRINITY_DN43845_c0_g1
      TRINITY_DN43845_c0_g1_i1

      TRINITY_DN43867_c0_g1
      TRINITY_DN43867_c0_g1_i1

      TRINITY_DN43843_c0_g1
      TRINITY_DN43843_c0_g1_i1

      TRINITY_DN43844_c0_g1
      TRINITY_DN43844_c0_g1_i1
```

Remarks

The de novo assembly part may take hours to days to proceed since this process in one of the time and resource-consuming in transcriptome data analysis. So that's OK if we couldn't accomplish the assembly as of the limited time for this workshop.

We have prepared the assembled transcripts for further analyses. You can use the following command to copy assembly files to your working directory.

```
# your current working directory: ~/Cpa_RNASeq/
cp /opt/Cpa_RNASeq/denovo_assembly/Trinity_2023-03-08.Trinity.fasta
```

Transcript Assembly Quality Assessment

According to suggestions from Trinity wiki of <u>transcriptome assembly quality</u> <u>assessment</u>, it's worth to determine how good of the quality of assembled transcript. Several approaches available for characterize the quality of your assembly. However, in this workshop we'll perform only two approaches

Examining gene and contig Nx statistics

We can compute Nx statistics from the assembled transcripts, as well as GC content, number of assembled transcripts, mean and median of contig length. from TrinityStats.pl command.

Activity

The following script in the Trinity toolkit will compute these values for you like so:

```
# Go to assembly directory
cd ~/Cpa_RNASeq/03_assembly
```

```
# Run TrinityStats
TrinityStats.pl Trinity_2023-03-08.Trinity.fasta
```

Expected time used: < 1 min

Expected output from the terminal screen:

```
Contig N10: 4225
Contig N20: 3083
Contig N30: 2464
Contig N40: 2014
Contig N50: 1678
```

Median contig length: 758 Average contig: 1106.61 Total assembled bases: 73890513

> Contig N10: 3987 Contig N20: 2871 Contig N30: 2280 Contig N40: 1862 Contig N50: 1531

Median contig length: 624 Average contig: 984.38 Total assembled bases: 48617582

The N10 through N50 values show the value of at least x% of number of assembled contigs have Nx nucleotide in length. For example, in contigs (isoform) level, the N50 indicates at least half (50%) of number of the assembled transcripts are 1,678 nucleotides in length, whereas N50 of the longest isoform that represent the gene is 1,531 nucleotides in length.

Benchmarking Universal Single-Copy Orthologs (BUSCO) analysis

BUSCO reported the transcriptome assembly completeness by evaluate whether the set of assembly recovered a whole set of universal functional genes referred from orthologous sequence from neighbor species. BUSCO metric is complementary to technical metrics like N50 as we did using TrinityStats.

Activity

BUSCO v4 and v5 use lineage datasets information from <u>OrthoDB</u> v10. You can search all available lineage datasets using the following command:

```
# Activate conda environment
conda activate busco
# List all lineage datasets in OrthoDB v10
busco --list-datasets
```

As of March 2023, more than 100 lineage datasets available in OrthoDB v10.

Generally the lineage to select for your assessments should be the most specific lineage available, e.g. for assessing Cyanophora transcriptome assembly data you may choose the Viridiplantae or Chlorophyta lineages rather than the metazoa lineage. Here we'll select Viridiplantae lineage dataset to evaluate the single-copy orthologs in Cyanophora assembled transcriptome using the following command.

```
# Go to current working directory
cd ~/Cpa_RNASeq/03_assembly
# Run BUSC0
busco --mode transcriptome \
--in Trinity_2023-03-08.Trinity.fasta \
--lineage_dataset /opt/Cpa_RNASeq/busco_downloads/lineages/viridipl
--out BUSCO_Viridiplantae \
--cpu 2 \
--offline
```

Estimated time usage: ~30-35 min

In this command, BUSCO runs in transcriptome mode by the required input file.

If we didn't make it in time, the backup data in backup BUSCO results will keep in: /opt/Cpa_RNASeq/BUSCO_Viridiplantae/, and BUSCO Viridiplantae lineage dataset is also in /opt/Cpa_RNASeq/busco_downloads/.

The classification results from BUSCO will save to BUSCO_viridiplantae in your working directory, as well as print out to the terminal as follow:

2023-03-08 00:05:05 INFO: Results: C:61.6% [s:42.8%,D:18.8%],F:16.0%,M:22.4%,n:425 2023-03-08 00:05:06 INFO:

> _____ |Results from dataset viridiplantae_odb10 T |C:61.6%[S:42.8%,D:18.8%],F:16.0%,M:22.4%,n:425 262 Complete BUSCOs (C) |182 Complete and single-copy BUSCOs (S) 1 Complete and duplicated BUSCOs (D) 80 68 Fragmented BUSCOs (F) T |95 Missing BUSCOs (M) 425 Total BUSCO groups searched I

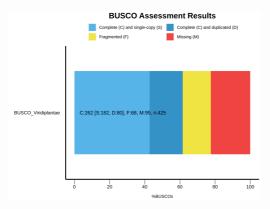
The BUSCO result shows the composition of the expected gene content within the assembled transcriptome. The BUSCO result can be divided into Complete and Single-Copy, Complete and Duplicated, Fragmented or Missing BUSCOs.

Activity

```
Now we'll generate a BUSCO plot.
# Working directory: cd ~/Cpa_RNASeq/03_assembly/assessment
generate_plot.py --working_directory BUSCO_viridiplantae
```

Estimated time usage: < 1 min

Expected graphical should be as follows. And explanation on each type of BUSCO results can be found at the <u>BUSCO's Documentation</u>.



Estimating Abundance and Differential Expression Analysis of Genes

Differential gene expression analysis is a statistical method that uses count data to determine significant changes between experimental groups. For example, transcriptional changes of stress-induced genes in plant leaves under water deficit compared to normal conditions are examined. Count data can be transcript, gene, exon, and noncoding characteristics.

To perform differential gene expression analysis with the Trinity integrated extensions, you will need the assembled transcripts/genes from the previous step and clean reads (and their replicates) from your experiment to assign to the assembled transcripts/genes and count the number of reads assigned to those transcripts/genes.

Estimating Transcript Abundance

This part will adopted from Trinity's Wiki Trinity Transcript Quantification.

There are two different methods for quantifying reads mapped to the reference, by using the alignment-based (RSEM) and alignment-free (salmon, kallisto) qualtifiers. In this workshop, we'll use the salmon, an ultra-fast alignment and quantification tool, to count number of reads mapped to the assembled gene.

Tip

You can see the usage of the command you wsh to perform by type the command followed by --help, or -h. For example, to see the usage of

align_and_estimate_abundance.pl -h

Activity

Now we will align and count reads mapped to the reference assembled transcripts using buit-in utility align_and_estimate_abundance.pl using the following commands.

1. Go to working directory and activate conda environment

Go to working directory

cd ~/Cpa_RNASeq

The working directory should contain the following subdirectories.

Cpa_RNASeq

├── 01_Rawdata ├── 02_QC ├── 03_assembly └── 04_DE_analysis └── 05_annotation

Then, activate conda environment

conda activate trinity

2. Estimating Transcript Abundance

your current working directory is: ~/Cpa_RNASeq

```
align_and_estimate_abundance.pl \
--transcripts 03_assembly/Trinity_2023-03-08.Trinity.fasta \
--seqType fq \
--samples_file /opt/Cpa_RNASeq/sample_list.tsv \
--est_method salmon \
--gene_trans_map 03_assembly/Trinity_2023-03-08.Trinity.fasta.gene_
--thread_count 2 \
--prep_reference
```

Then, type 1s to see the results in your working directory. These directories are the results from the above command. Each folder represents each biological replicate in your experiment.

```
drwxrwxr-x 5 jiratchaya jiratchaya 4096 Mar 9 20:37 dark_1/
drwxrwxr-x 5 jiratchaya jiratchaya 4096 Mar 9 20:38 dark_2/
drwxrwxr-x 5 jiratchaya jiratchaya 4096 Mar 9 20:38 dark_3/
drwxrwxr-x 5 jiratchaya jiratchaya 4096 Mar 9 20:38
normal_light_1/
drwxrwxr-x 5 jiratchaya jiratchaya 4096 Mar 9 20:39
normal_light_2/
drwxrwxr-x 5 jiratchaya jiratchaya 4096 Mar 9 20:39
normal_light_3
```

Estimated time usage: ~ 20 min per user

Parameter descriptions of align_and_estimate_abundance.pl: the assembled transcripts flagged in --transcripts, --seqType indicates file format of reads that will be mappped to the reference transcripts. We define the read count estimation tool in --est_method, in this workshop we use salmon, and also estimate read counts using information of gene-transcript relationships from the trinity_out_dir.Trinity.fasta.gene_trans_map file that we specified in the --gene_trans_map parameter.

A list of read files will be contained in the metadata file sample_list.tsv in the parameter --samples_file, which we have prepared for you. In short, the sample list will be prepared in a tab-delimited text file indicating the relationships between biological replicates. For example,

```
jiratchaya@pslab1:~$ cat /opt/Cpa_RNASeq/sample_list.tsv
dark
       dark_1 /opt/Cpa_RNASeq/Cyanophora_rawdata/SRR8306034_1.fastq
/opt/Cpa_RNASeq/Cyanophora_rawdata/SRR8306034_2.fastq
       dark_2 /opt/Cpa_RNASeq/Cyanophora_rawdata/SRR8306029_1.fastq
dark
/opt/Cpa_RNASeq/Cyanophora_rawdata/SRR8306029_2.fastq
dark
       dark_3 /opt/Cpa_RNASeq/Cyanophora_rawdata/SRR8306028_1.fastq
/opt/Cpa_RNASeq/Cyanophora_rawdata/SRR8306028_2.fastq
normal_light
               normal_light_1
/opt/Cpa_RNASeq/Cyanophora_rawdata/SRR8306033_1.fastq
/opt/Cpa_RNASeq/Cyanophora_rawdata/SRR8306033_2.fastq
normal_light
               normal_light_2
/opt/Cpa_RNASeq/Cyanophora_rawdata/SRR8306032_1.fastq
/opt/Cpa_RNASeq/Cyanophora_rawdata/SRR8306032_2.fastq
normal_light
               normal_light_3
/opt/Cpa_RNASeq/Cyanophora_rawdata/SRR8306035_1.fastq
/opt/Cpa_RNASeq/Cyanophora_rawdata/SRR8306035_2.fastq
```

The first column indicates the study experimental groups, followed by their biological replicates in the second column, and the forward and reverse sequence read files belong to their biological replicate. It's important that the file path begins with the directory in which you'll be working so that the programs can correctly route to the files.

Output results are created in the current working directory separately for experimental groups and biological replicates as follow.

dark_1

```
├── aux_info/
    ├── cmd_info.json
    ├─ lib_format_counts.json
    ├── libParams/
    ├-- logs/
    ├─ quant.sf
    └── quant.sf.genes
normal_light_3
    ├── aux_info/
    ├── cmd_info.json
    ├── lib_format_counts.json
    ├── libParams/
    ├-- 1ogs/
    ├-- quant.sf
    └── quant.sf.genes
     Activity
```

According to the previous part, now we'll organize the directory to make it tidy by moving all the results to the directory 04_DE_analysis.

```
# make sure you're in ~/Cpa_RNASeq directory so that you can move 1
mv dark* normal* 04_DE_analysis/
# Then enter to the directory `04_DE_analysis`
cd 04_DE_analysis
ls ./*
```

Expected result:

```
(trinity) jiratchaya@pslab1:~/Cpa_RNASeq/04_DE_analysis$ ls ./*
./dark_1:
aux_info cmd_info.json lib_format_counts.json libParams logs
quant.sf quant.sf.genes
./dark_2:
aux_info cmd_info.json lib_format_counts.json libParams logs
quant.sf quant.sf.genes
./dark_3:
aux_info cmd_info.json lib_format_counts.json libParams logs
quant.sf quant.sf.genes
./normal_light_1:
aux_info cmd_info.json lib_format_counts.json libParams logs
quant.sf quant.sf.genes
./normal_light_2:
aux_info cmd_info.json lib_format_counts.json libParams logs
quant.sf quant.sf.genes
./normal_light_3:
aux_info cmd_info.json lib_format_counts.json libParams logs
quant.sf quant.sf.genes
```

after running salmon you'll find output files:

- quant.sf: transcript abundance estimates (generated by salmon)
- quant.sf.genes: gene-level abundance estimates (generated here by summing transcript values)

Here's an example of quant.sf.genes file:

Building Transcript and Gene Expression Matrices

We'll estimates abundance matrices with the filename quant.sf, which are available in all results directories. In this step, the utility abundance_estimates_to_matrix.pl is used to combine all separate count matrices from the file quant.sf in all result directories into a single matrix file. By using salmon as --est_method and specifying the parameter --gene_trans_map, a gene abundance matrix is created.

Activity

1. Create abundance matrix

Your current working directory: ~/Cpa_RNASeq/04_DE_analysis

```
abundance_estimates_to_matrix.pl --est_method salmon \
--gene_trans_map ../03_assembly/Trinity_2023-03-08.Trinity.fasta.ge
--name_sample_by_basedir \
*/quant.sf
```

Expected result:

```
(trinity) jiratchaya@pslab1:~/Cpa_RNASeq/04_DE_analysis$ ls -l
total 4945
drwxrwx--- 1 root PSLab 4096 Mar 5 16:16 dark_1
drwxrwx--- 1 root PSLab 4096 Mar 5 16:16 dark_2
drwxrwx--- 1 root PSLab 4096 Mar 5 16:16 dark_3
drwxrwx--- 1 root PSLab 4096 Mar 5 16:16 normal_light_1
drwxrwx--- 1 root PSLab 4096 Mar 5 16:16 normal_light_2
drwxrwx--- 1 root PSLab 4096 Mar 5 16:16 normal_light_3
-rw-rw---- 1 root PSLab 67 Mar 6 13:29
salmon.gene.counts.matrix
-rw-rw---- 1 root PSLab 67 Mar 6 13:29
salmon.gene.TPM.not_cross_norm
-rw-rw---- 1 root PSLab 2736773 Mar 6 13:29
salmon.isoform.counts.matrix
-rw-rw---- 1 root PSLab 2297101 Mar 6 13:29
salmon.isoform.TPM.not_cross_norm
```

This command will generate 4 result files:

- salmon.gene.counts.matrix is the estimated raw RNA-Seq counts in GENE level in all experimental groups.
- salmon.gene.TPM.not_cross_norm is the Transcript per Million (TPM) of RNA-Seq counts in GENE level in all experimental groups.
- salmon.isoform.counts.matrix is the estimated raw RNA-Seq counts in TRANSCRIPTS level in all experimental groups.
- salmon.isoform.TPM.not_cross_norm is the Transcript per Million (TPM) of RNA-Seq counts in TRANSCRIPTS level in all experimental groups.

Quality Control of Sample Read Counts and Biological Replicates

Once you've performed quantification for each experimental group, it's good to examine the data to ensure that your biological replicates are well correlated, and also to investigate relationships among your samples. It is critical that you identify any obvious differences between the relationships between your sample and replicates, such as those resulting from accidental mislabeling of sample

replicates, strong outliers, or batch effects, prior to further data analysis. The Trinity's utility called Ptr (pronounced as 'Peter', stands for Perl to R) can generate some exploratory data analysis rely on count matrix, such as compare difference between replicate, compare difference between experimental groups, principal component analysis, and so on.

Activity

Recheck the current working directory

pwd

You must be in ~/Cpa_RNASeq/04_DE_analysis

TheThen prepare the sample metadata from differential expression analysis (DE). The sample metadata table for the DE analysis is different from the table used for abundance estimation. We only need the first two columns from this file to create a metadata table for the analysis of DE. Therefore, we can use the following Bash command to create and edit a new file.

EExtract the first 2 columns of metadata to estimate the read count into a new file in 04_DE_analysis

```
cut -f 1-2 /opt/Cpa_RNASeq/sample_list.tsv > samples.txt
```

See expected result file

(trinity) jiratchaya@pslab1:~/Cpa_RNASeq/04_DE_analysis\$ cat samples.txt dark dark_1 dark dark_2 dark dark_3 normal_light normal_light_1 normal_light normal_light_2 normal_light normal_light_3

Compare replicates for each of your samples

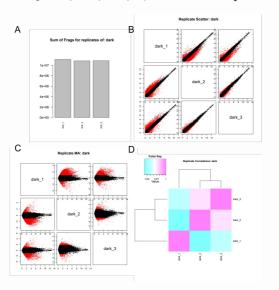
This step will use PtR to reads the matrix of counts, performs a counts-permillion (CPM) data transformation followed by a log2 transform, and then generates a multi-page pdf file named <code>\${sample}.rep_compare.pdf</code> for each of your samples, including several useful plots

Activity Compare replicates for each of your samples # Current workdir: ~/Cpa_RNASeq/04_DE_analysis PtR --matrix salmon.isoform.counts.matrix \ --samples samples.txt --log2 --CPM \ --min_rowSums 10 \ --compare_replicates

These files will append more to your current working directories:

```
-rw-rw---- 1 root PSLab 4695 Mar 6 14:37
salmon.isoform.counts.matrix.R
-rw-rw---- 1 root PSLab 1990182 Mar 6 14:37 dark.rep_compare.pdf
-rw-rw---- 1 root PSLab 1828692 Mar 6 14:37
normal_light.rep_compare.pdf
-rw-rw---- 1 root PSLab 3558147 Mar 6 14:37
salmon.isoform.counts.matrix.minRow10.CPM.log2.dat
```

The last 3 files are newly generated by this step. There's two PDF files separated by experimental groups, dark.rep_compare.pdf and normal_light.rep_compare.pdf, and raw data for plots in .dat file.



Example result of comparing biological replicates in Dark samples. The figures were captured from dark.rep_compare.pdf file. (A) The sum of mapped fragments. (B) Pairwise comparisons of replicate log(CPM) values, in which the data points more than 2-fold different are highlighted in red. (C) The pairwise MA plots (x-axis: mean log(CPM), y-axis log(fold_change)). And, (D) A Replicate Pearson correlation heatmap.

Compare Replicates Across Samples

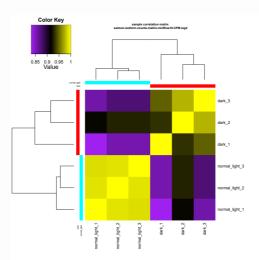
Activity

This command will generate a useful heatmap of pearson correlation matrix of samples from two different experimental groups.

```
PtR --matrix salmon.isoform.counts.matrix \
--min_rowSums 10 \
-s samples.txt \
--log2 --CPM \
--sample_cor_matrix
```

These files will append more to your current working directories:

```
-rw-rw---- 1 root PSLab 4012 Mar 6 15:02
salmon.isoform.counts.matrix.R
-rw-rw---- 1 root PSLab 3558147 Mar 6 15:02
salmon.isoform.counts.matrix.minRow10.CPM.log2.dat
-rw-rw---- 1 root PSLab 678 Mar 6 15:02
salmon.isoform.counts.matrix.minRow10.CPM.log2.sample_cor.dat
-rw-rw---- 1 root PSLab 6429 Mar 6 15:02
salmon.isoform.counts.matrix.minRow10.CPM.log2.sample_cor_matrix.pdf
```



heatmap of pearson correlation coefficiant between Dark and Normal light samples.

Principal Component Analysis (PCA)

Another important analysis method to explore relationships among the sample replicates is Principal Component Analysis (PCA).

You can find more explanation about PCA here: https://blog.bioturing.com/2018/06/14/principal-component-analysisexplained-simply/ - https://youtu.be/FgakZw6K1QQ

Activity
PtR --matrix salmon.isoform.counts.matrix \
-s samples.txt \
--min_rowSums 10 --log2 \

--CPM --center_rows \

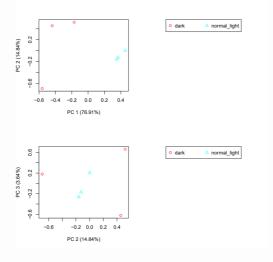
--prin_comp 3

These files will append more to your current working directories:

```
-rw-rw---- 1 root PSLab
                          4789 Mar 6 15:18
salmon.isoform.counts.matrix.R
-rw-rw---- 1 root PSLab 4069112 Mar 6 15:18
salmon.isoform.counts.matrix.minRow10.CPM.log2.centered.dat
-rw-rw---- 1 root PSLab
                           756 Mar 6 15:18
salmon.isoform.counts.matrix.minRow10.CPM.log2.centered.PCA.prcomp.score
S
-rw-rw---- 1 root PSLab 4163653 Mar 6 15:18
salmon.isoform.counts.matrix.minRow10.CPM.log2.centered.PCA.prcomp.loadi
s
-rw-rw---- 1 root PSLab
                           5446 Mar 6 15:18
salmon.isoform.counts.matrix.minRow10.CPM.log2.centered.prcomp.principal
f
4
                                                                     ►
```

You can find the PCA plot in

salmon.isoform.counts.matrix.minRow10.CPM.log2.centered.prcomp.principal_components.prin



PCA plot.

We set the number of principal components (PC) to be calculated for only first 3 PCs in --prin_comp. Which indicates that these PCs will be plotted, as shown above, with PC1 vs. PC2 and PC2 vs. PC3. In this example, the replicates cluster tightly according to sample type, which is very reassuring.

Differential Expression Analysis

Trinity also contains a built-in utility for DE analysis called run_DE_analysis.pl, in which use the count matrix and sample metadata file. Trinity provides support for several differential expression analysis tools, currently including edgeR, DESeq2, limma/voom, and ROTS.

DE analysis in Trinity will perform pairwise comparison of gene/transcript expression. If the biological replicates are presented for each sample, you should indicate this as we already created in our metadata table samples.txt. Here we'll analyze DE genes in the 'transcript' level using the salmon.isoform.counts.matrix file.

Activity

DE analysis using DESeq2

```
run_DE_analysis.pl \
--matrix salmon.isoform.counts.matrix \
--method DESeq2 \
--samples_file samples.txt \
--output DESeq2_result
```

After run the above command, the following directory will append to your current working directory:

drwxrwx--- 1 root PSLab 688 Mar 6 15:42 DESeq2_result

In this output directory, you'll find the following files for each of the pairwise comparisons performed:

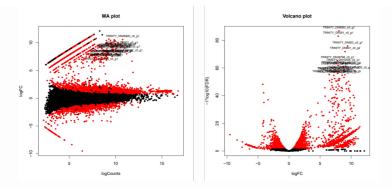
```
-rw-rw---- 1 root PSLab 972633 Mar 6 15:42
salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.count_matrix
-rw-rw---- 1 root PSLab 4247784 Mar 6 15:42
salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results
-rw-rw---- 1 root PSLab 2428272 Mar 6 15:42
salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results.MA_n
f
-rw-rw---- 1 root PSLab 1845 Mar 6 15:42
salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.Rscript
```

Result explanations:

- salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.Rscript is the R-script executed to perform the DE analysis.
- salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.count_matrix is an integer matrix of read count derived from the input file salmon.isoform.counts.matrix.
- salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results is the DE analysis results, including log fold change and statistical significance.
- salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results.MA_n_Volcano. is MA and Volcano plots features found DE at the defined FDR will be colored red.

Here's an example of DE analysis result file (salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results):

An example of volcano plot for transcript-level differentially expression analysis.



(left) MA plot and (right) volcano plot.

Extracting and clustering differentially expressed transcripts

An initial step in analyzing differential expression is to extract those transcripts that are most differentially expressed (most significant FDR and fold-changes) and to cluster the transcripts according to their patterns of differential expression across the samples.

Activity

Extracting and clustering differentially expressed transcripts can run using the following from within the DE output directory, by running the following script:

```
cd DESeq2_result/
analyze_diff_expr.pl \
--matrix ../salmon.isoform.counts.matrix \
-P 1e-3 \
-C 2 \
--samples ../samples.txt \
--max_genes_clust 10000
```

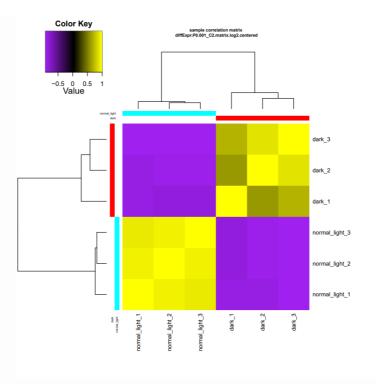
The above command use an integer count matrix from DE analysis, and define criteria for extracting differentially expressed transcripts. For example, set p-value cutoff for FDR in -P to 0.001, set minimum absolute log 2-fold change criteria in -C to 2, meaning that it will extracted only the DE transcripts that are $2^2 = 4$ -fold, and use only top 10,000 among all differentially transcripts in --max_genes_clust for hierarchical clustering analysis. However, user can customize these criteria based on their interest.

The following results will append to the current working directory DESeq2_result

```
-rw-rw---- 1 root PSLab 120 Mar 6 16:34
salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results.samp
S
-rw-rw---- 1 root PSLab 332012 Mar 6 16:34
salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results.P0.0
-UP.subset
-rw-rw---- 1 root PSLab 42038 Mar 6 16:34
salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results.P0.0
-UP.subset
-rw-rw---- 1 root PSLab 373901 Mar 6 16:34
salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results.P0.0
t
-rw-rw---- 1 root PSLab
                           51 Mar 6 16:34
DE_feature_counts.P0.001_C2.matrix
-rw-rw---- 1 root PSLab 73959 Mar 6 16:34 diffExpr.P0.001_C2.matrix
-rw-rw---- 1 root PSLab 4649 Mar 6 16:34
diffExpr.P0.001_C2.matrix.R
-rw-rw---- 1 root PSLab 246973 Mar 6 16:34
diffExpr.P0.001_C2.matrix.log2.centered.dat
-rw-rw---- 1 root PSLab
                          698 Mar 6 16:34
diffExpr.P0.001_C2.matrix.log2.centered.sample_cor.dat
                         6399 Mar 6 16:34
-rw-rw---- 1 root PSLab
diffExpr.P0.001_C2.matrix.log2.centered.sample_cor_matrix.pdf
-rw-rw---- 1 root PSLab 101250 Mar 6 16:34
diffExpr.P0.001_C2.matrix.log2.centered.genes_vs_samples_heatmap.pdf
-rw-rw---- 1 root PSLab 14777602 Mar 6 16:34
diffExpr.P0.001_C2.matrix.RData
4
                                                                   F
```

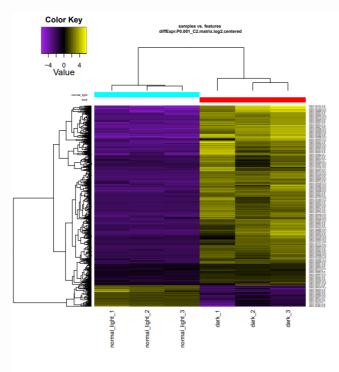
Result explanations:

- salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results.samples is identical to the metadata samples.txt file.
- salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results.P0.001_C2.dar
 UP.subset is the subset of expression matrix of up-regulated transcripts in
 Dark group, which are down-regulated in Normal light group.
- salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results.P0.001_C2.nor UP.subset is the subset of expression matrix of up-regulated transcripts in Normal light group, which are down-regulated in Dark group.
- salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results.P0.001_C2.DE. is a summary of DE transcripts results containing columns of significant values, and its normalized expression value.
- diffExpr.P0.001_C2.matrix.log2.centered.sample_cor_matrix.pdf is the sample correlation matrix, as follow.



Sample correlation matrix visualized only for differentially expressed transcripts.

• diffExpr.P0.001_C2.matrix.log2.centered.genes_vs_samples_heatmap.pdf is heatmap of differentially expressed transcripts.



Heatmap of differentially expressed transcripts.

DE gene patterning and clustering analysis

In the heat map of differentially expressed transcripts, there is a clear difference between the DE transcripts under dark and normal light conditions. Therefore, using define_clusters_by_cutting_tree.pl, we can divide these genes into clusters based on the same trend of expression values as follows.

Activity

Automatically Partitioning Genes into Expression Clusters

```
define_clusters_by_cutting_tree.pl \
  -R diffExpr.*.matrix.RData \
  --Ptree 60
```

There are three different methods for dividing genes into clusters, K-Means clustering, hierarchical clustering (as used in the heatmap), and the recommended method of using criteria to truncate tree branch lengths that fall below the criteria by using '–Ptree'.

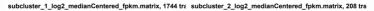
The following results will append to the current working directory DESeq2_result which are files and diffExpr.P0.001_C2.matrix.RData.clusters_fixed_P_60/ directory.

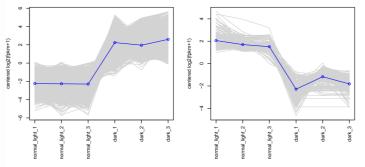
```
-rw-rw---- 1 root PSLab 45837 Mar 6 16:51
clusters_fixed_P_60.heatmap.heatmap_gene_order.txt
-rw-rw---- 1 root PSLab 61467 Mar 6 16:51
clusters_fixed_P_60.heatmap.gene_cluster_colors.dat
-rw-rw---- 1 root PSLab 110890 Mar 6 16:51
clusters_fixed_P_60.heatmap.heatmap.pdf
drwxrwx--- 1 root PSLab 400 Mar 6 16:51
diffExpr.P0.001_C2.matrix.RData.clusters_fixed_P_60/
```

List of files in diffExpr.P0.001_C2.matrix.RData.clusters_fixed_P_60/ directory are:

-rw-rw---- 1 root PSLab 43915 Mar 6 16:51 my_cluster_plots.pdf
-rw-rw---- 1 root PSLab 220780 Mar 6 16:51
subcluster_1_log2_medianCentered_fpkm.matrix
-rw-rw---- 1 root PSLab 26259 Mar 6 16:51
subcluster_2_log2_medianCentered_fpkm.matrix
-rw-rw---- 1 root PSLab 816 Mar 6 16:51 __tmp_plot_clusters.R

The DE transcript partiitoning and clustering is located in my_cluster_plots.pdf





DE transcript partiitoning and clustering analysis

Then, we'll subset the assembled transcriptome for only differentially expressed genes for functional annotation analysis in the next chapter.

Activity

From the previous command, we already have a list of differentialli expressed gnees in

salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results.P1e-3_c2.DE.subset file. For the functional annotation analysis in the next step, we will subset only top 10 upregulated and top 10 downregulated DEGs for annotation step. Data subsetting will use the following command.

1. Extract top 10 upregulated DEGs in Normal light condition. This command will ascendingly sort the 6th column (log2foldchange) of the file

salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results.P1e-3_C2.normal_light-UP.subset, then selected top 10 highest log2foldchange from the last 10 lines, then select only the trinity transcript ID in the first column.



 Do the same for dark condiiton using salmon.isoform.counts.matrix.dark_vs_normal_light.DESeq2.DE_results.P1e-3_C2.dark-UP.subset file.

sort --key=6 --numeric-sort salmon.isoform.counts.matrix.dark_vs_nc

3. Concatenate transcript ID files (DEGtop10_NormalLight-up.txt and DEGtop10_dark-up.txt) into a single file.

cat DEGtop10_NormalLight-up.txt DEGtop10_dark-up.txt > DEGtop20_all

4. Retrieve FASTA sequence of top 20 DEGs using Trinity's utility.

retrieve_sequences_from_fasta.pl DEGtop100_all.txt ~/Cpa_RNASeq/03_

The output file locates in ~/Cpa_RNASeq/04_DE_analysis/DESeq2_result

Transcriptome Assembly Annotation

In the absence of a well-annotated reference genome of the species of your choice, researchers will need to generate a draft genome or transcriptome through a de novo approach. Subsequent analyses after assessing the completeness of the assembly include differential gene expression analysis (quantitative approach), functional annotation (qualitative), and many others depending on your experimental design. Functional annotation involves searching for the biological meaning of the biological sequence of interest using known data sets or predicting a new one.

Functional annotation of non-model organisms is usually compared with neighboring species that have a well-annotated reference genome in terms of functions. For example, complete genome/transcriptome datasets of the black tiger shrimp Penaeus monodon or the Caridian shrimp *Marsupenaeus japonicus* can be used as reference datasets for annotating the transcriptomes of the banana shrimp *Fenneropenaeus merguiensis*. Information from the taxonomic rank, e.g., a complete genome/transcriptome of the phylum Arthropoda or subphylum Crustacea, can also be used to annotate the banana shrimp dataset. In addition, orthologous data from the associated kingdom, such as the eukaryote dataset, can be used to annotate the banana shrimp dataset.

Functional annotation is task to find the biological meaning such as biochemical and biological function of proteins or CDS. Possible analyses to annotate genes can be for example:

- Sequence similarity searching: such as <u>BLAST</u>, <u>HMMER</u>, <u>Diamond</u>, and many more. This approach uses biological sequences of interest (proteins or nucleotides) as query sequences to search a database of known sequence annotations (called subject sequences) and check how similar your query is to the subject sequence.
- Gene Ontology (GO) annotation: is functional classification of genes into 3 major classes; cellular component, biological process, and molecular function. More information please see <u>Gene Ontology Overview</u>.
- **Biological pathway and interaction network:** by searching the orthologous protein from pathway and biological interaction network such as <u>KEGG pathway</u>, <u>STRING</u>, <u>Reactome</u>, and many more. This approach provide information of the gene/protein of interest interacting with others within the same pathway of gene set.

Functional Annotation using EggNOG-mapper

EggNOG-mapper is a tool for functional annotation analysis in biological sequences, which can be proteomes, genomes, transcriptomes or metagenomes. EggNOG-mapper uses the EggNOG database as a reference for searching orthologous identifiers. Many search algorithms have been deployed to search against EggNOG database, such as diamond (fast and comparable), MMseqs2 fast and comparable, and HMMER3 (slowest).

Once user have already installed EggNog-mapper, you will need to download

EggNOG database to your local machine using the following command. But in this workshop, we already prepared it for you :)

download_eggnog_data.py -D -M -y -f --data_dir eggnog_db/

The following command download_eggnog_data.py tries to download the diamond search database by default, but in this workshop we will use the MMSeqs2 search tool, so we can skip downloading the diamond database with -D and download the MMSeqs2 database using -M instead.

Activity

After preparing the EggNOG database, we can run the EggNog mapper via the file emapper.py as in the following command, specifying the required arguments for the path of the downloaded EggNOG database in -- data_dir, the input type (--itype) and the search algorithm MMSeqs2 (-m). You can see usage of emapper.py by type emapper.py -help in your terminal, or have a look at <u>A few recipes of using EggNOG-mapper</u>.

```
# Go to main project directory
cd ~/Cpa_RNASeq
# Activate conda environment
conda activate emapper
# Run EggNOG-mapper
emapper.py --cpu 5 \
--data_dir /opt/Cpa_RNASeq/eggNOG_db/ \
--output_dir 05_annotation/ \
-m diamond \
--evalue 1e-5 \
-i 04_DE_analysis/DESeq2_result/DEGtop100_all_seqs.fasta \
--no_file_comments \
--itype CDS \
--excel \
--output Cpa_emapper_2023-03-10
```

Estimated time usage: ~ 1 hr

Expected output files:

```
-rw-r--r- 1 jiratchaya jiratchaya 20287555 Mar 8 22:54
Cpa_emapper_2023-03-10.emapper.annotations
-rw-r--r- 1 jiratchaya jiratchaya 5060493 Mar 8 22:54
Cpa_emapper_2023-03-10.emapper.annotations.xlsx
-rw-r--r- 1 jiratchaya jiratchaya 5333975 Mar 8 22:54
Cpa_emapper_2023-03-10.emapper.hits
-rw-r--r- 1 jiratchaya jiratchaya 2528455 Mar 8 22:54
Cpa_emapper_2023-03-10.emapper.seed_orthologs
```

Output explanations:

- *.emapper.annotations is a result file from the annotation phase. Each row represents the annotation reported for a given query.
- *.emapper.annotations.xlsx is as same as *.emapper.annotations but in Excel format.

- *.emapper.hits is a result file from the MMseqs2 search phase.
- .emapper.seed_orthologs is a result file from parsing the hits. Each line associates a query with a seed ortholog. This file has the same format regardless of which searcher was used, except that it can be in short format (4 fields) or full format.

Homology Searching using NCBI BLAST

BLAST (Basic Local Alignment Search Tools) is usually a first choice as sa sequence similarity search tool. BLAST search for region that similar to both of our sequence of interest, which can be nucleotides or proteins. BLAST can be used to infer functional and evolutionary relationships between sequences as well as help identify members of gene families.



Several types of BLAST search algorithms classified by type of query sequence (protein/nucleotide sequence you want to search for) and th expected subject sequence (protein/nucleotide in BLAST database expected to match with the query sequence) such as the following table

Search algorithms	Input type	Output type
BLASTN	Nucleotide	Nucleotide
BLASTX	CDS	Protein
TBLASTN	Protein	CDS
BLASTP	Protein	Protein

- **BLASTN (Nucleotide BLAST)** compares nucleotide sequences to nucleotide sequences in databases such as Nt, 16S rRNA, ITS, and custom nucleotide databases. A useful example of using BLASTN is to search for phylogenetic relationships between the query sequence and neighboring species and infer the evolutionary relationship between them.
- **BLASTX (Translated BLAST)** uses the nucleotide query sequence to search protein databases such as Nr, Uniprot, Protein Databank (PDB), and custom protein databases. BLASTX translates the query into six reading frames (-3, -2, -1, 1, 2, 3) before searching. Therefore, the time required for BLASTX is about 6 times slower than the straightforward BLAST. A useful example of using BLASTX is to search for possible translation frames in de novo transcriptome assembly.
- **TBLASTN (Translated BLAST)** uses protein query sequence to search nucleotide databases by translating into six reading frames. TBLASTN provides benefit in searching for coding sequence (CDS) along with its open reading frame from protein sequence.
- BLASTP (Protein BLAST) compares protein query sequences with protein

sequences in databases. A useful example of using BLASTP is to search for protein sequence similarities to infer their functions from conserved domains observed in the sequence.

Activity

In this workshop, we' will use the nucleotide database BLAST using custom nucleotide data. In short, we'll subdivide whole nucleotide sequences and create the database BLAST from the NCBI nucleotide collection database belonging to the phylum Chlorophyta, Rhodophyta and Glaucophyta. **We have already prepared these databases for you** by using the following command:

makeblastdb -in [input_seq.fasta] -dbtype [nucl|prot]

Database paths:

- Chlorophyta BLASTN database: /opt/Cpa_RNASeq/BLAST_DB/Chlorophyta.fna
- Rhodophyta BLASTN database: /opt/Cpa_RNASeq/BLAST_DB/Rhodophyta.fna
- Glaucophyta BLASTN database: /opt/Cpa_RNASeq/BLAST_DB/Glaucophyta.fna

Depending on your interest, you can choose which phylum you want to use as BLAST database as above. Then we' will investigate it together!

BLASTN

BLAST tools is located in the ncbi environment, so you will need to activate environment as follow:

conda activate ncbi

Then, run BLASTN:

```
# Current working directory: ~/Cpa_RNASeq
blastn -db /opt/Cpa_RNASeq/BLAST_DB/[database_of_interest] \
-query 04_DE_analysis/DEG_sequence.fasta \
-out 05_annotation/BLASTN_DEG_[phylum_of_interest].tsv \
-evalue 1e-5 \
-outfmt "7 std qcovhsp stitle" \
-max_target_seqs 5 \
-num_threads 4
```

Estimated time: < 5 mins

By this command, you'll search query sequence in your database of interest. The result will collected using BLAST if E-value <= 1e-5.

Here we' specify the output format 7, which results in the table file containing the comment lines that start with the '#' character.

Example BLASTN output format 7 std qcovhsp stitle:

```
$ head -n 15 BLASTN_DEG_Chlorophyta.tsv
# BLASTN 2.13.0+
# Query: sampleA
# Database: /opt/Cpa_RNASeq/BLASTN/Chlorophyta.fna
# 0 hits found
# BLASTN 2.13.0+
# Query: TRINITY_DN281_c0_g1_i1
# Database: /opt/Cpa_RNASeq/BLASTN/Chlorophyta.fna
# Fields: query acc.ver, subject acc.ver, % identity, alignment length,
mismatches, gap opens, q. start, q. end, s. start, s. end, evalue, bit
score, % query coverage per hsp, subject title
# 10 hits found
TRINITY_DN281_c0_g1_i1 XM_043068042.1 99.797 1972
                                                      0
                                                              1
1
       1968
               2126
                     155
                              0.0
                                      3616
                                              100
                                                      XM_043068042.1
Chlamydomonas reinhardtii uncharacterized protein (CHLRE_12g498600v5),
mRNA
                                                              5
TRINITY_DN281_c0_g1_i1 DQ122889.1
                                      94.995 1918
                                                      84
       1913
               1936
                       26
                               0.0
                                      3000
                                              97
                                                      DQ122889.1
1
Chlamydomonas incerta elongation factor alpha-like protein (efl) mRNA,
complete cds
TRINITY_DN281_c0_g1_i1 XM_001696516.2 98.929 1400
                                                      15
                                                              0
       1913 1539
                      140
                               0.0
                                      2503
                                              71
514
                                                      XM 001696516.2
Chlamydomonas reinhardtii uncharacterized protein (CHLRE_06g263450v5),
mRNA
TRINITY_DN281_c0_g1_i1 XM_043062829.1 98.929 1400
                                                      15
                                                              0
514
       1913 1539
                       140
                               0.0
                                      2503
                                              71
                                                      XM_043062829.1
Chlamydomonas reinhardtii uncharacterized protein (CHLRE_06g263450v5),
mRNA
TRINITY_DN281_c0_g1_i1 CP097822.1
                                      100.000 1261
                                                      0
                                                              0
       1261
             3287383 3286123 0.0
                                      2329
                                              64
                                                      CP097822.1
1
Chlamydomonas reinhardtii strain CC-5816 chromosome 12
TRINITY_DN281_c0_g1_i1 CP097822.1
                                      100.000 251
                                                              0
                                                      0
       1510
               3285932 3285682 2.34e-128
                                              464
1260
                                                      13
CP097822.1 Chlamydomonas reinhardtii strain CC-5816 chromosome 12
```

The result file can be further open with spreadsheet software such as Microsoft Excel, by skipping the comment lines. The column name is described in # Fields: comment line, containing the following fields:

Field names	Descriptions
query acc.ver	Query sequence ID
subject acc.ver	Subject sequence ID
% identity	Percentage identity
alignment length	Alignment length
mismatches	Number of mismatches
gap opens	Number of gap openings
q. start	Query sequence alignment start position
q. end	Query sequence alignment end position
s. start	Subject sequence alignment start position
s. end	Subject sequence alignment end position
evalue	Expect value
bit score	Bit score
% query coverage per hsp	Query Coverage Per High Scoring Pairs
subject title	Subject sequence name

You can specify options to include in this tabular format by type the following command and take a look at the terminal.

blastn <mark>-help</mark>

Reference sources

- De novo transcriptome assembly, annotation, and differential expression analysis from Galaxy Training!
- NCBI Bioinformatics Resources: <u>An Introduction: BLAST: Compare &</u> <u>identify sequences</u>. Berkeley Library, University of California.

References

- Buffalo, Vince. 2015. *Bioinformatics Data Skills: Reproducible and Robust Research with Open Source Tools.* " O'Reilly Media, Inc.".
- Grabherr, Manfred G, Brian J Haas, Moran Yassour, Joshua Z Levin, Dawn A Thompson, Ido Amit, Xian Adiconis, Lin Fan, Raktima Raychowdhury, and Qiandong Zeng. 2011. "Trinity: Reconstructing a Full-Length Transcriptome Without a Genome from RNA-Seq Data." *Nature Biotechnology* 29 (7): 644.
- Haas, Brian J, Alexie Papanicolaou, Moran Yassour, Manfred Grabherr, Philip D Blood, Joshua Bowden, Matthew Brian Couger, David Eccles, Bo Li, and Matthias Lieber. 2013. "De Novo Transcript Sequence Reconstruction from RNA-Seq Using the Trinity Platform for Reference Generation and Analysis." *Nature Protocols* 8 (8): 1494–1512.

Knopp, Michael, Sriram G. Garg, Maria Handrich, and Sven B. Gould. 2020.
"Major Changes in Plastid Protein Import and the Origin of the Chloroplastida." *iScience* 23 (3): 100896. <u>https://doi.org/10.1016/j.isci.2020.100896</u>.

Martin, Jeffrey A., and Zhong Wang. 2011. "Next-Generation Transcriptome

Assembly." *Nature Reviews Genetics* 12 (10): 671–82. https://doi.org/10.1038/nrg3068.