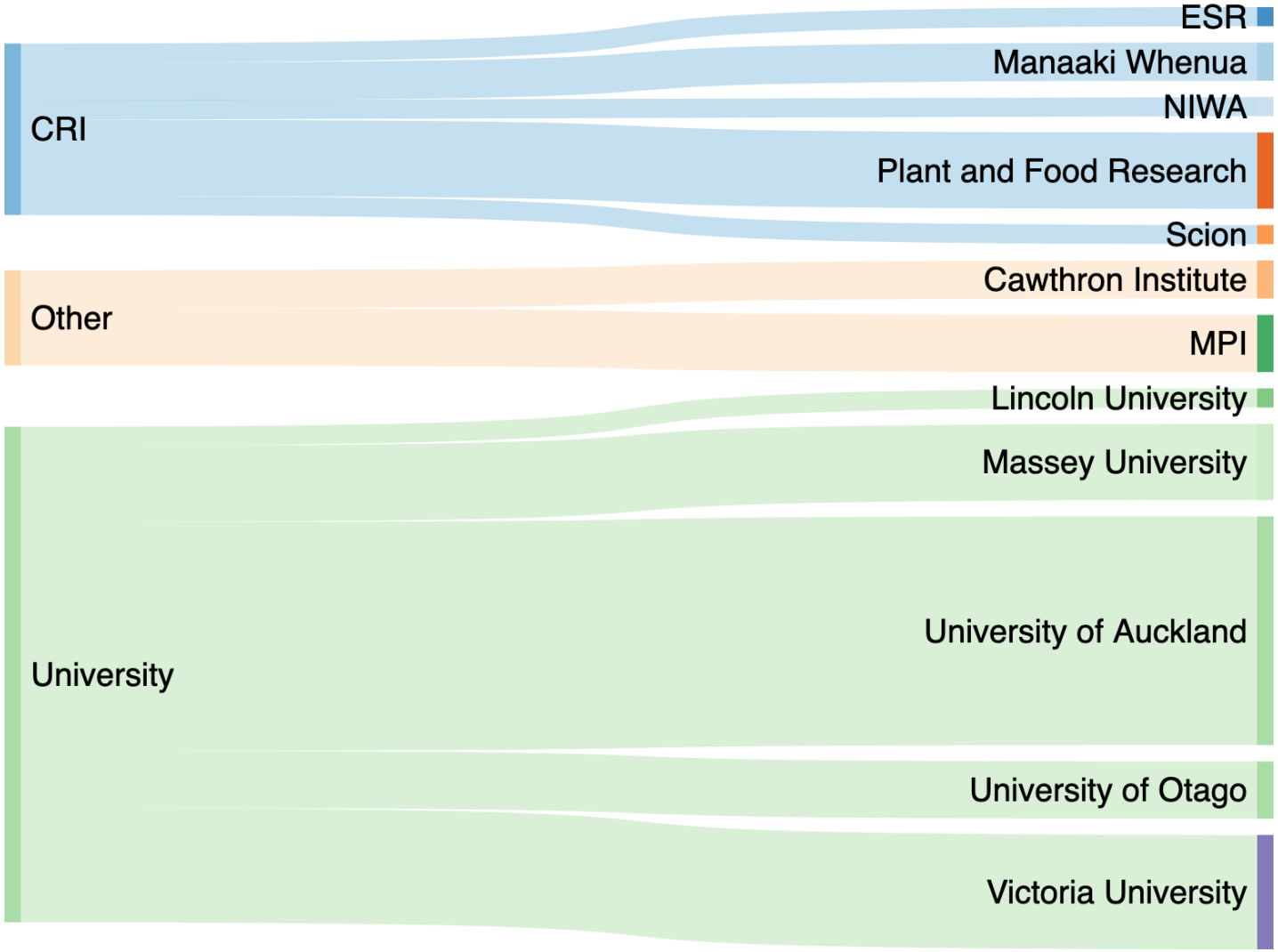
genomics aotearoa

Metagenomics Summer School 2020

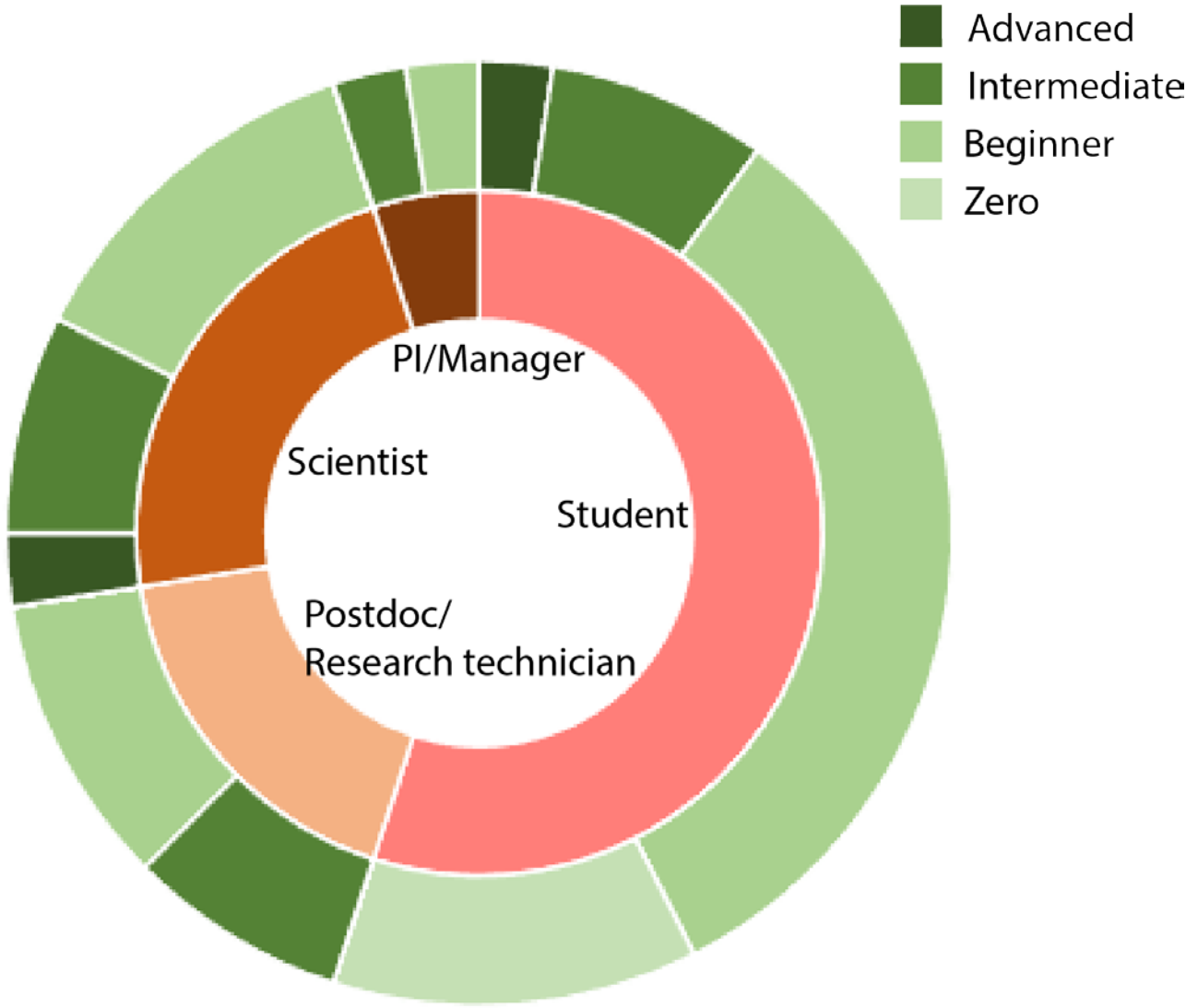
Welcome!

- **Housekeeping**
- **Overview of attendees**
  - Where are we from?
  - How experienced are we?
- **Any questions?**

Where are we from?



How experienced are we?



Genomics Aotearoa - Resources

Genomics Aotearoa – GitHub repositories

https://github.com/GenomicsAotearoa/

- Metagenomics Summer School material
- RNA seq workshop
- Environmental metagenomics
  - Metagenomic annotation and binning
- Methods and musings
  - Bin cluster refinement
  - Genome assembly ont
  - Metagenomic ont

Structure of workshop

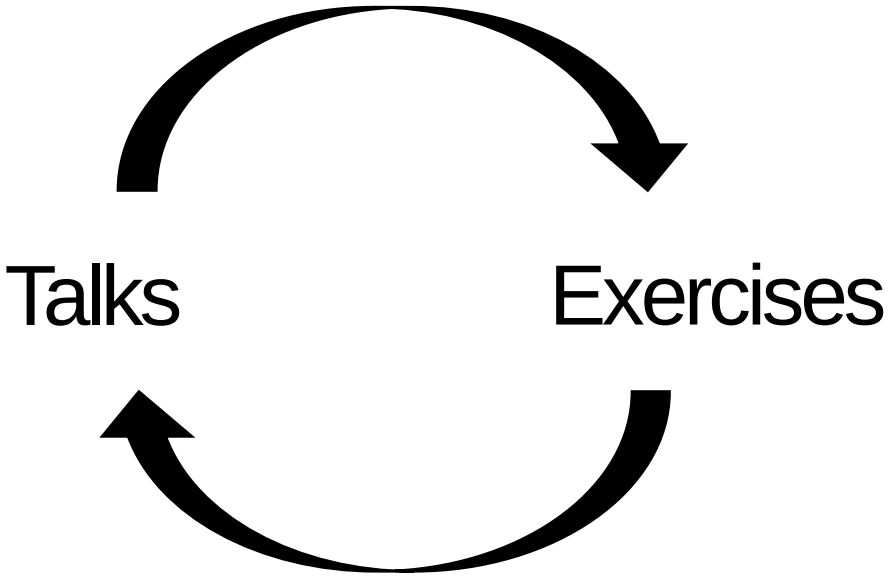
Workshop exercises

Day 1

1. Bash scripting
2. Quality filtering raw reads
3. Assembly (part 1)
4. Assembly (part 2)
5. Evaluating the assembly

Day 2

1. Binning (part 1, read mapping)
2. Binning (part 2, initial binning)
3. Binning (part 3, dereplication)
4. Bin refinement



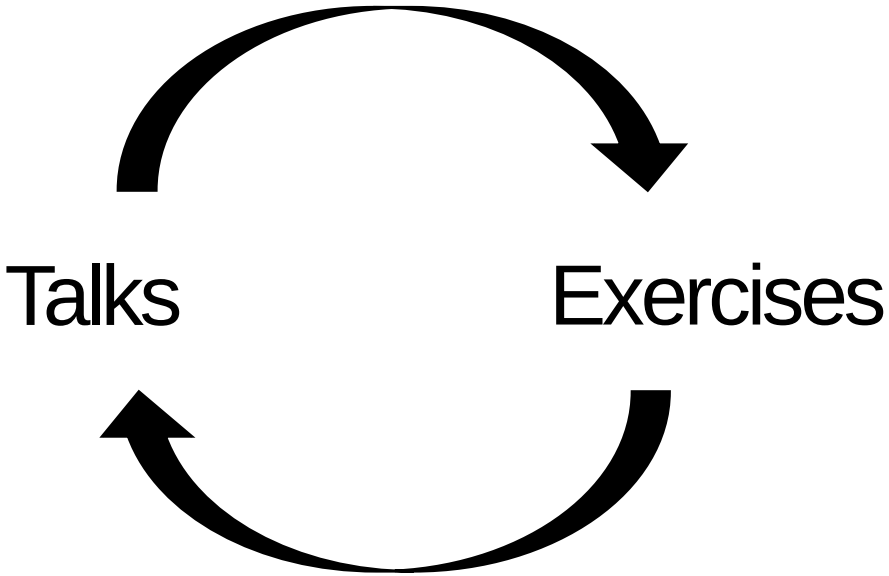
Structure of workshop

Day 3

1. Viruses
2. Coverage and Taxonomy
3. Gene prediction
4. Gene annotation (part 1)
5. Gene annotation (part 2)

Day 4

1. Gene annotation (part 3)
2. Presentation of data



Starting each session

1. Log in to the NeSI Jupyter hub via a browser
2. Also log in to NeSI via a standard terminal
3. Open the workshop exercise materials on GitHub
4. *Optional: Open a text editor such as Notepad or Jupyter text editor*

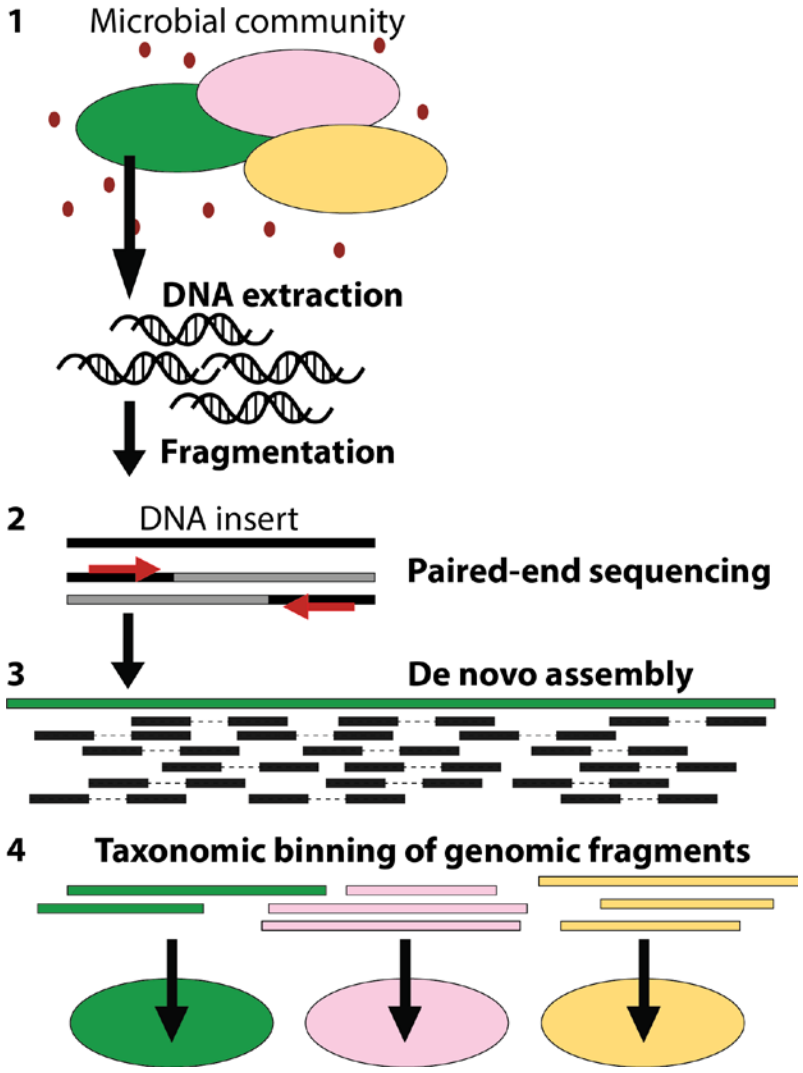
Day 1

Introduction to Bash scripting
Decision tree
Quality filtering WGS data
Genome assembly

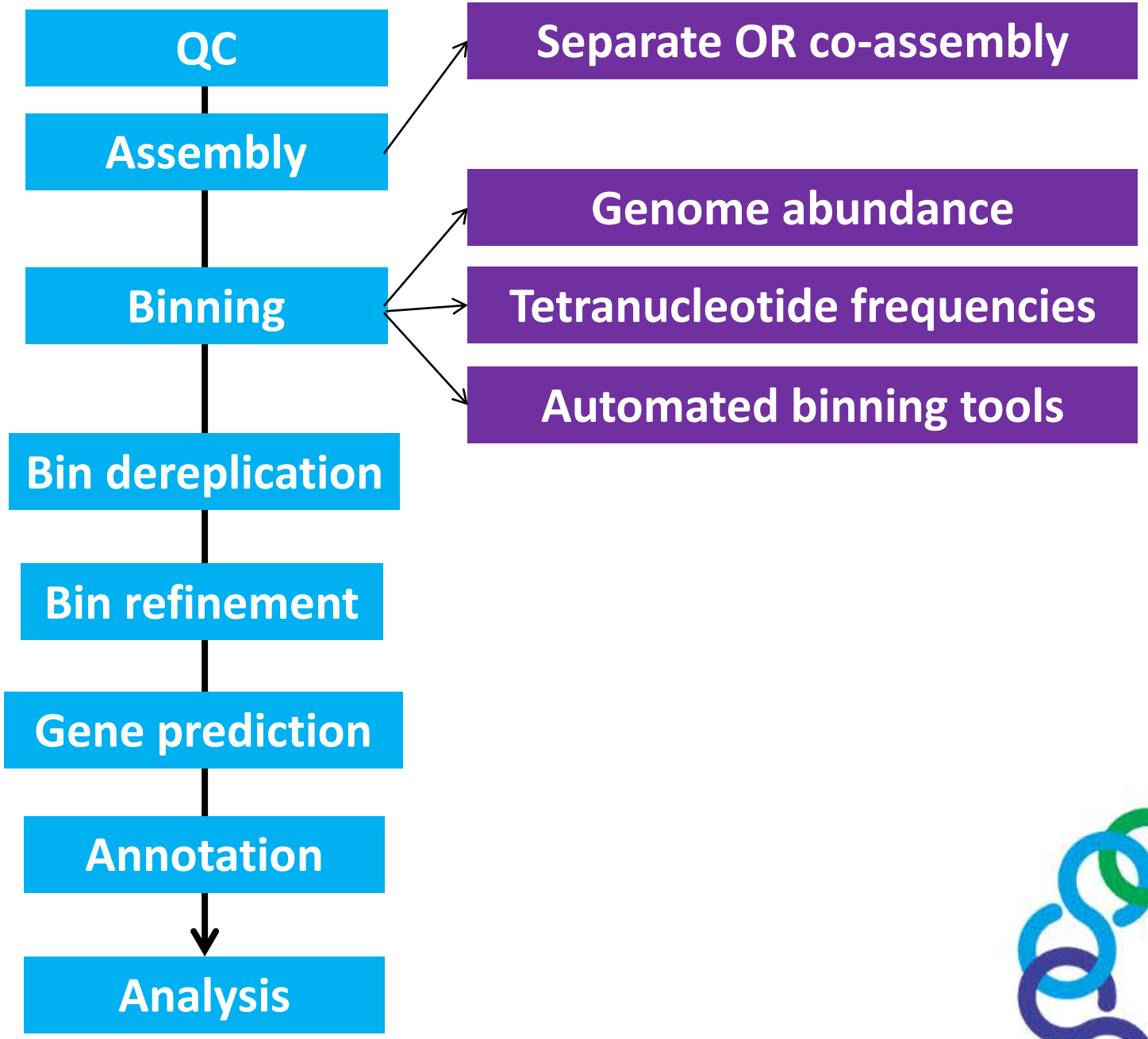
Bash scripting

Metagenomic decision tree(s)

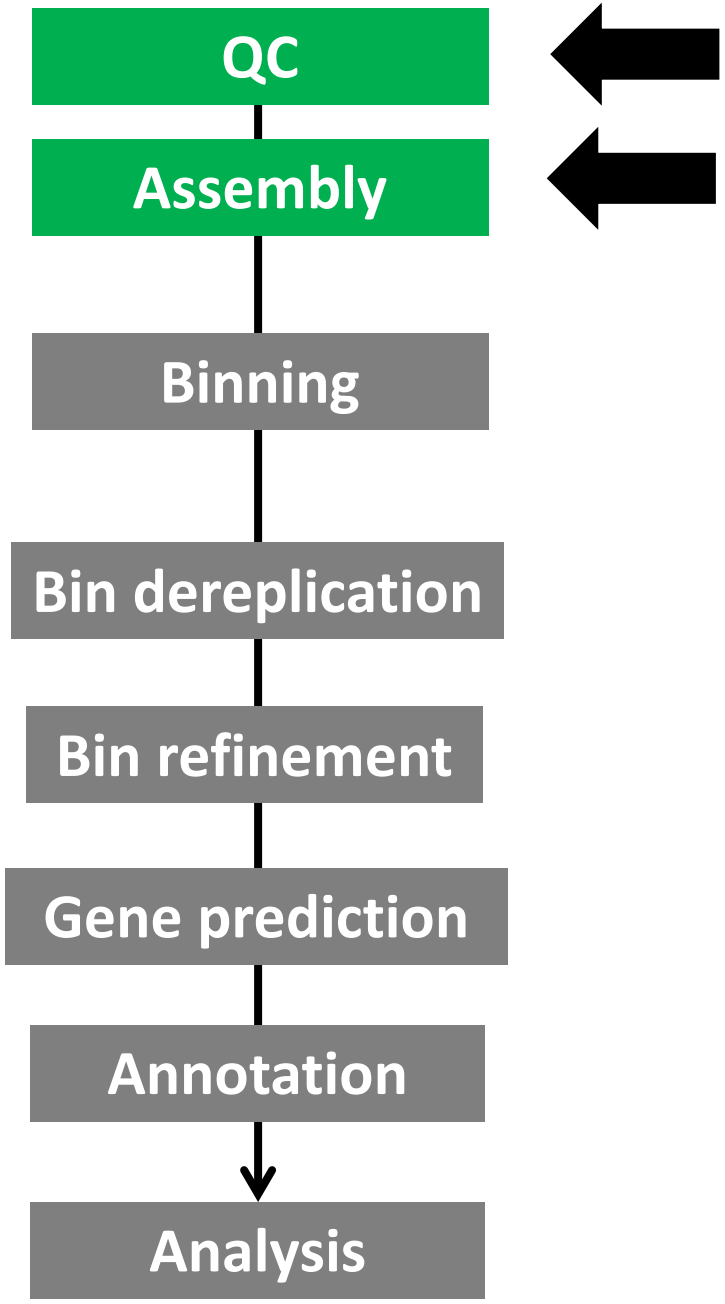
Our goal: genome recovery



Overview


QC
Assembly
Binning
Bin dereplication
Bin refinement
Gene prediction
Annotation
Analysis
Separate OR co-assembly
Genome abundance
Tetranucleotide frequencies
Automated binning tools

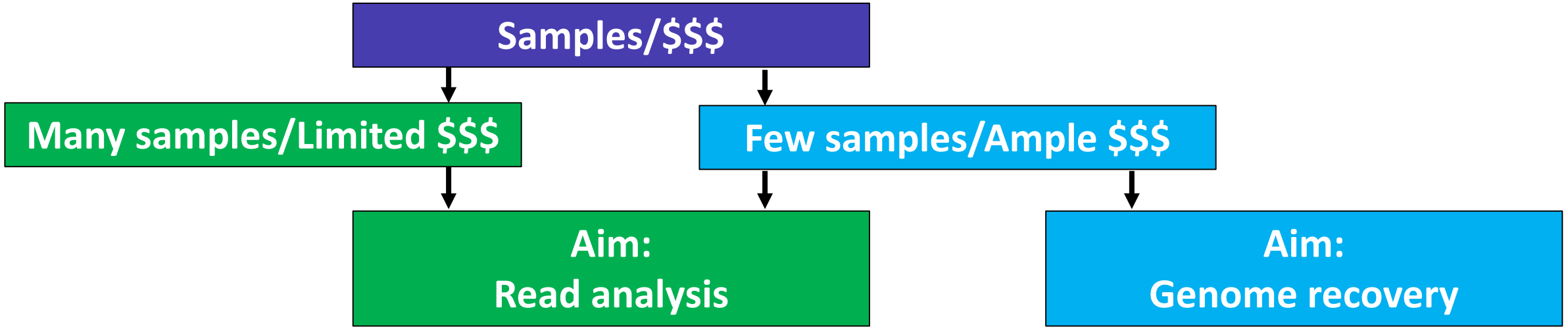
Overview


QC
Assembly
Binning
Bin dereplication
Bin refinement
Gene prediction
Annotation
Analysis

Decision tree

- **Starts with experimental design**
- **DNA extraction**
- **WGS library prep**
- **Amount of sequencing**


Samples/$$$
Many samples/Limited $$$
Few samples/Ample $$$


Samples/$$$
Many samples/Limited $$$
Few samples/Ample $$$
Aim:
Read analysis
Aim:
Genome recovery

Samples/$$$

Many samples/Limited $$$

Few samples/Ample $$$

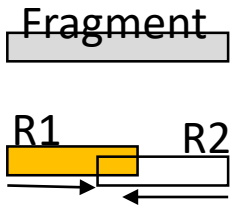
Aim:
Read analysis

Aim:
Genome recovery

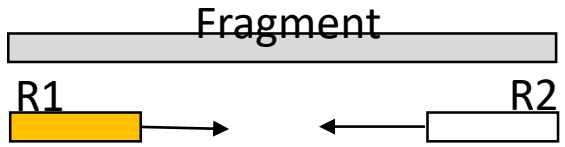
Libraries:
Short overlapping PE inserts

Libraries:
Longer gapped PE inserts

(e.g. 200 bp DNA fragments
for 2x125 bp reads)

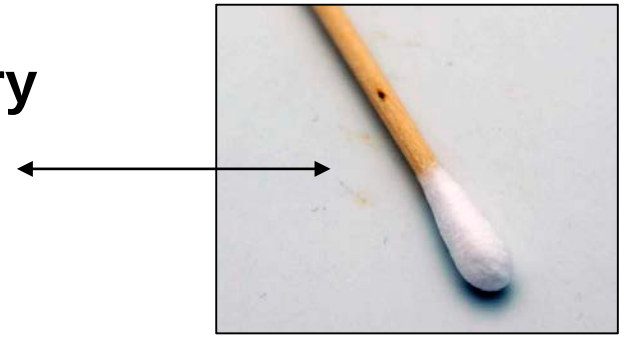


(e.g. >=550 bp DNA fragments)



DNA input

- **Very low inputs (e.g. nanograms) for Nextera library prep = enzymatic fragmentation with broad size distributions**



- **High inputs (e.g. 100s ng) for TruSeq = physical fragmentation with defined size selection**

Tends to yield sequences of larger inserts

Samples/$$$

Many samples/Limited $$$

Few samples/Ample $$$

Aim:
Read analysis

Aim:
Genome recovery

Libraries:
Short overlapping PE inserts

Libraries:
Longer gapped PE inserts

(e.g. 200 bp DNA fragments
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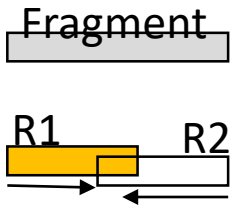

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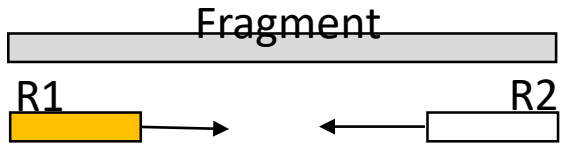


Samples/$$$

Many samples/Limited $$$

Few samples/Ample $$$

Aim:
Read analysis

Aim:
Genome recovery

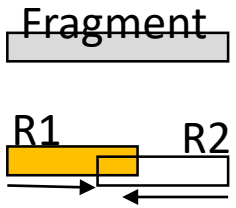
Libraries:
Short overlapping PE inserts

Libraries:
Longer gapped PE inserts

Sequencing depth:
Shallow (<10 Gbp)

Sequencing depth:
Deep (e.g. >=10s Gbp)

(e.g. 200 bp DNA fragments
for 2x125 bp reads)



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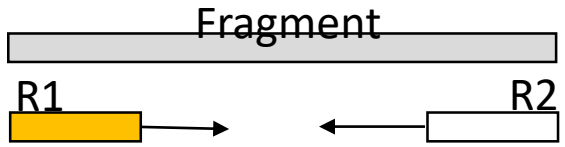
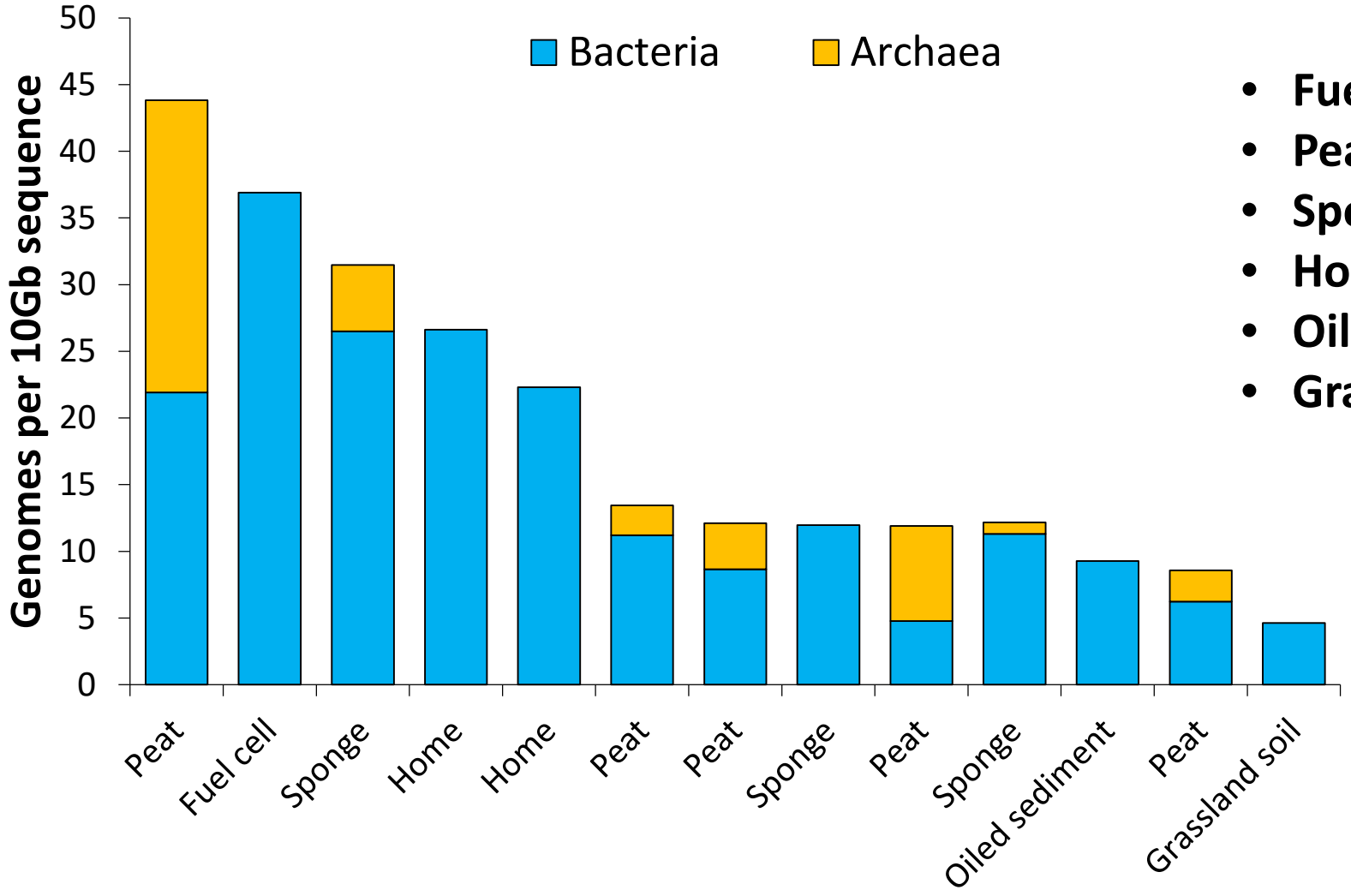
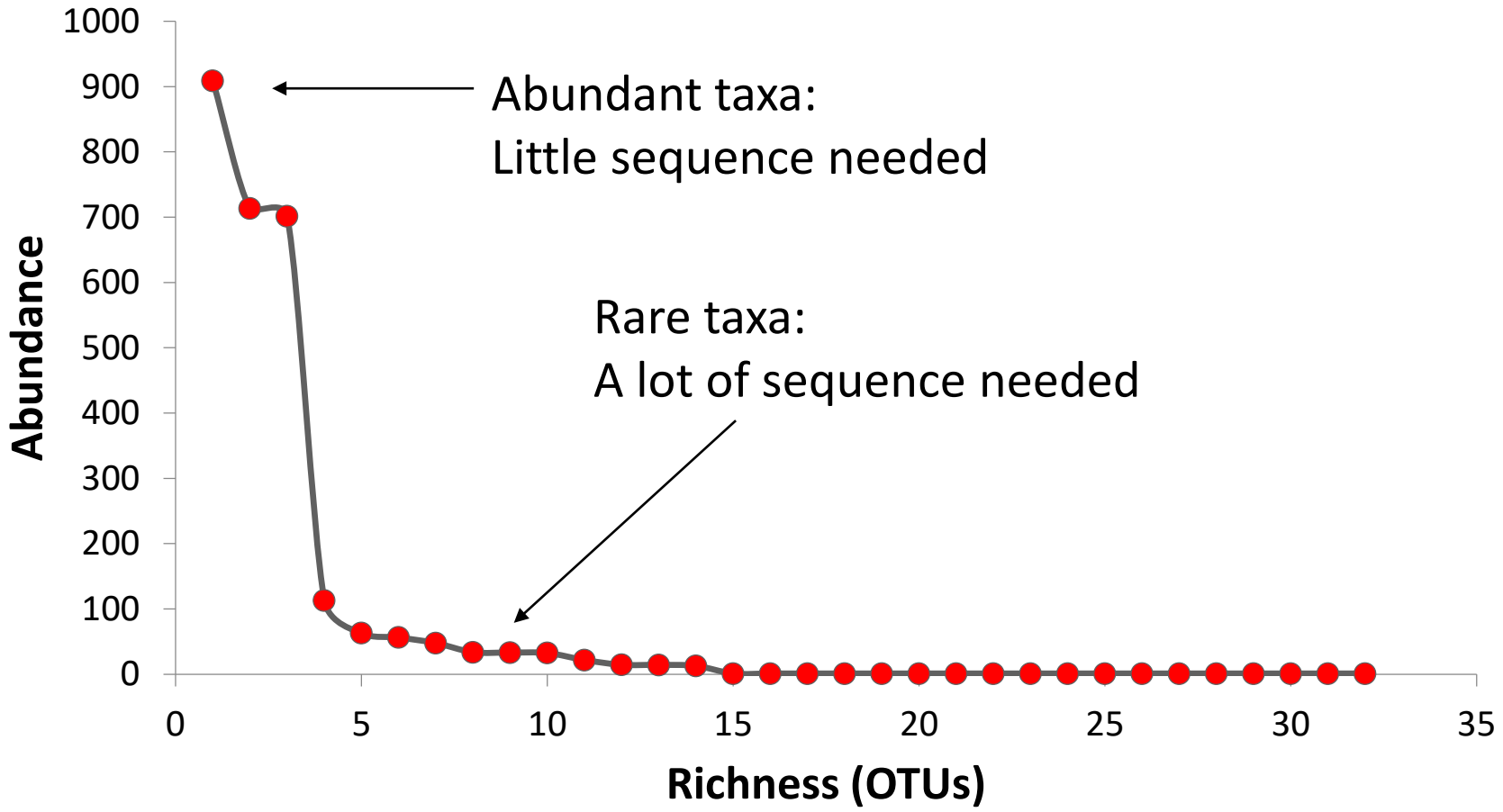


Genome recovery per environment

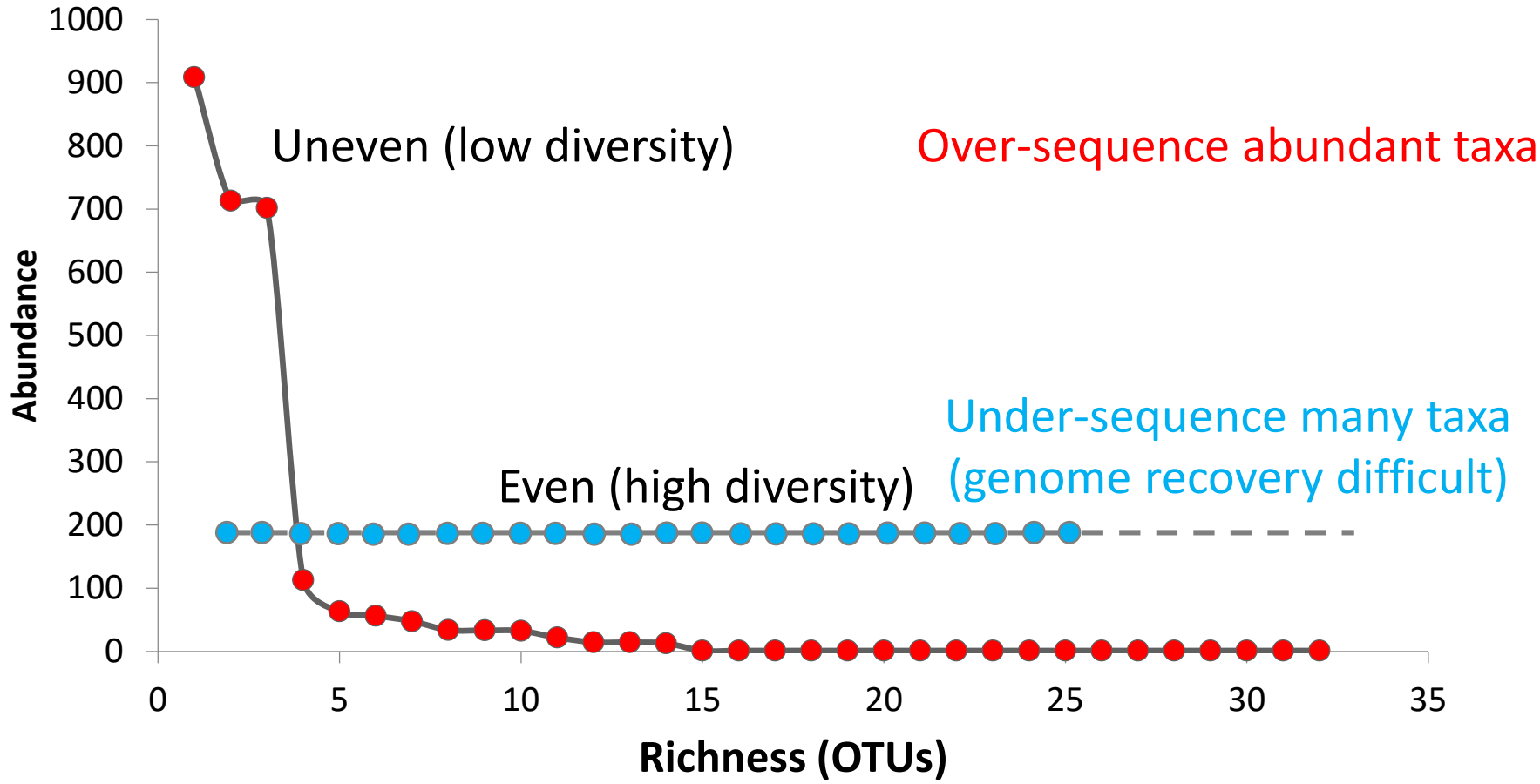


- **Fuel cell microbiome**
- **Peat (boreal)**
- **Sponge microbiome**
- **Home microbiome**
- **Oiled sediment (seafloor)**
- **Grassland soil**

Estimate sequencing depth

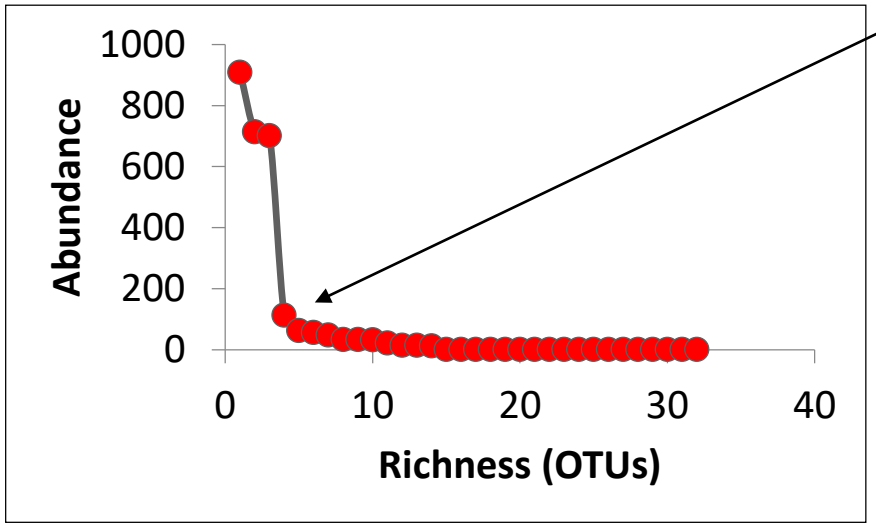


Community structure matters



Estimate sequencing depth

- **Estimate generously**
- **Determine/guesstimate relative abundance of rarest target organism**
- **Determine/guesstimate the average genome size**
- **Factor in larger eukaryote genomes**
- **Decide the minimum desired coverage (e.g. 30x)**



e.g., 5% relative abundance = 5% of sequence data

Prokaryotic genome sizes

Estimate sequencing depth

- Estimate generously
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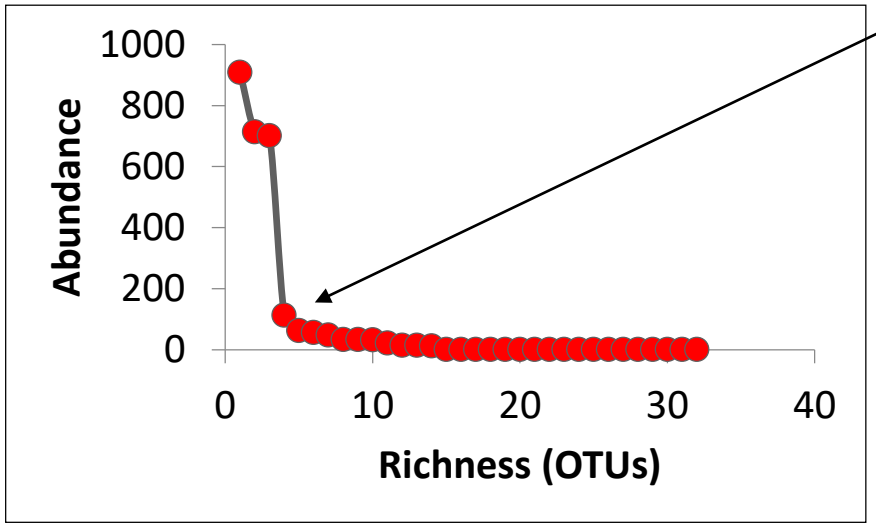
- Determine/guesstimate the average genome size
- Factor in larger eukaryote genomes
- Decide the minimum desired coverage (e.g. 30x)



e.g., 5% relative abundance = 5% of sequence data

Mock parameters:

- Bacterial genome 5 Mbp long
- 5% abundance (need 100/5 or 20x a genome at 100% abundance)
- 30x coverage

5 Mbp x 20 x 30 = 3,000 Mbp (or 3 Gbp)

When you have so many genomes

You need a:

- **Clear goal**
- **Question**
- **Hypothesis to test**

Quality control/filtering

The FastQ data format

```
@SEQUENCE_1
ATCGATCGATCG
+
4:<AIIIFIIII
@SEQUENCE_2
AATGATCCATG
+
IIIIIIIIIII
@SEQUENCE_3
TGTGTGACATG
+
BBGBBCIFIII
```

Each sequence is represent by four lines

1. Sequence name
2. Sequence content
3. Spacer line (+, or +Sequence name)
4. Quality information

The FastQ data format

- **What does the quality score even mean?**
  - **It represents the probability of a nucleotide position being *<u>incorrectly</u>* called**

$$Q = -10\ log_{10}p$$

Q	p	Prob. correct
0	1	0
10	0.1	0.9
20	0.01	0.99
30	0.001	0.999
40	0.0001	0.9999

How each Q value is encoded varies between sequencing platforms

<u>Generally</u> we work with the **Illumina 1.8+** (Phred+33) standard

Quality filtering WGS data

- Remove barcode and adapter regions
- Remove low-quality regions of reads
- Identify potential problems during sequencing
  - Deciphering ‘aberrant’ metrics in FastQC
  - Adapter read-through

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Task: Quality filtering

Visualising data with FastQC

1. Inspecting *fastq* files
2. Identifying regions of concern

Quality filtering with trimmomatic

1. Removing adapter sequences
2. Removing low-quality regions

Common issues with WGS data

Do I need to remove adapters?	→	Yes.
I don't know if adapters have been removed or not	→	Check the per-nucleotide distributions You will see 100% skews if they remain.
What's the lowest Q to allow when trimming?	→	Assembly is a self-correcting process, so you can be surprisingly lenient
What if my GC skew is outside of the expected range?	→	FastQC is calibrated to genome data where you expect GC conservation. Metagenomes do not adhere to this assumption

How to interpret over-represented kmers

What do high sequence duplication levels mean?

Issues when working with low-input DNA libraries

Discussion: Filtering out host DNA

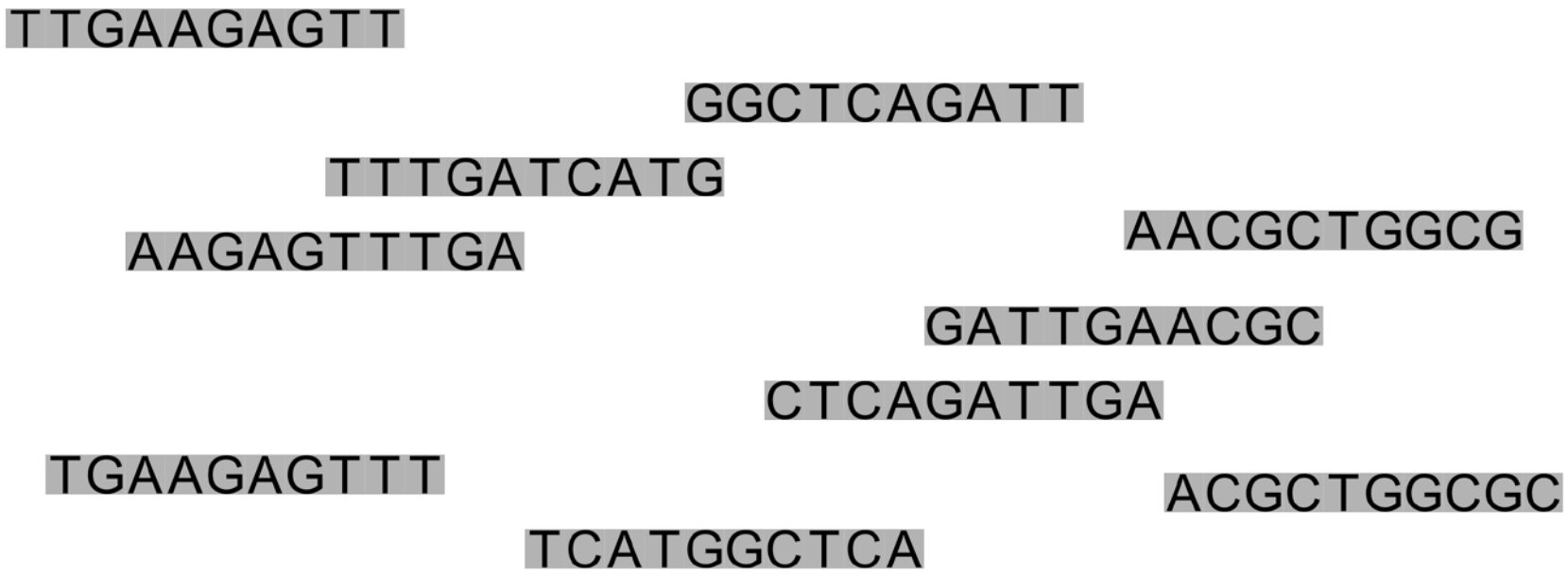
Assembly

Genome assembly

Overlap-Consensus-Layout (OCL) assembly

Genome assembly

Overlap-Consensus-Layout (OCL) assembly



Genome assembly

Overlap-Consensus-Layout (OCL) assembly

The problem for *de novo* assembly?

$$N.comparisons = \frac{(n)(n-1)}{2} = \frac{(10)(10-1)}{2} = 45$$

Genome assembly

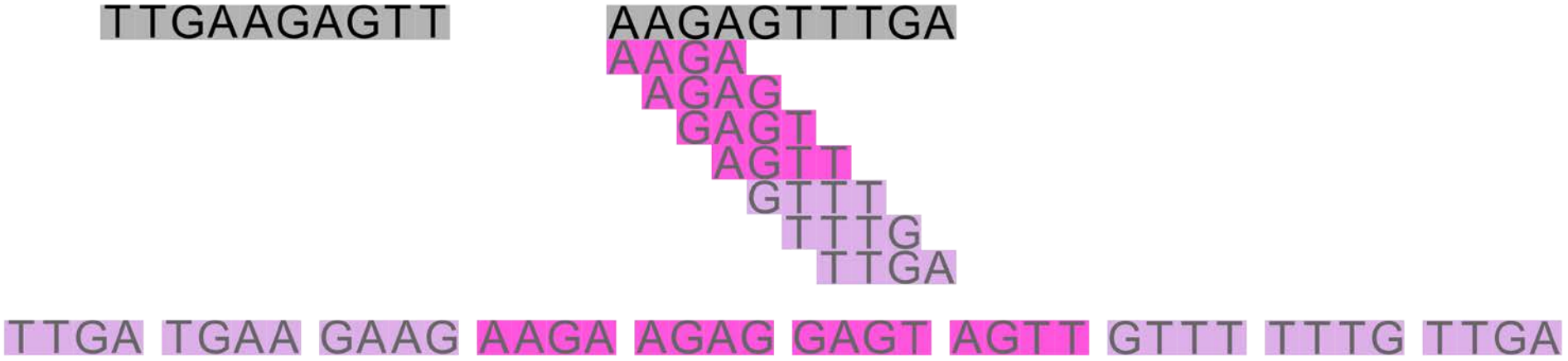
De Bruijn graph assembly

Break reads into shorter *k*-mers

Genome assembly

De Bruijn graph assembly

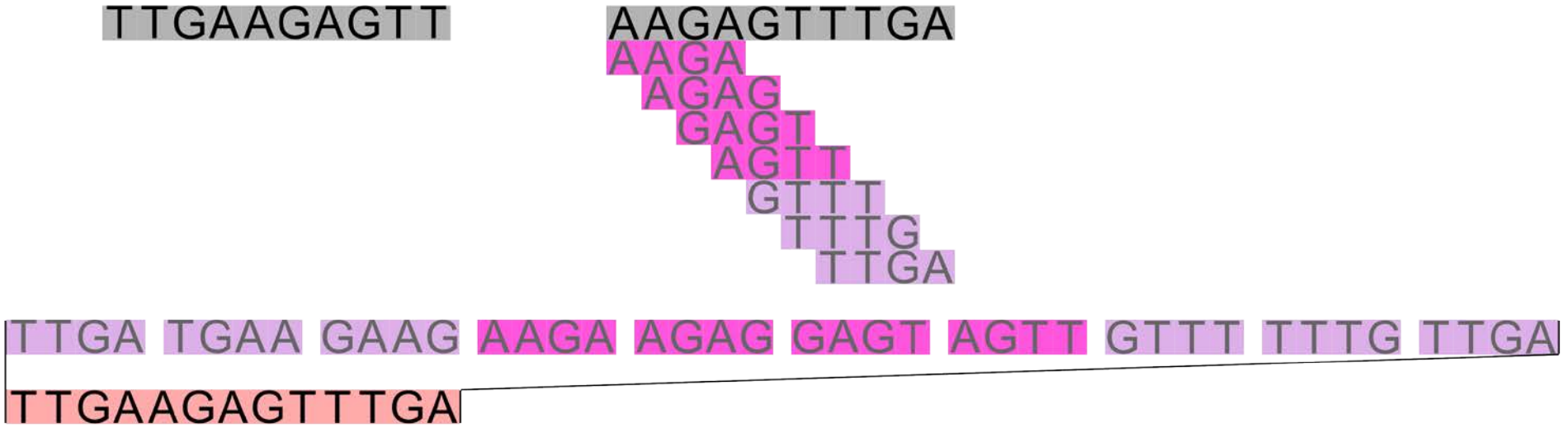
Identify sequences of shared *k*-mers



Genome assembly

De Bruijn graph assembly

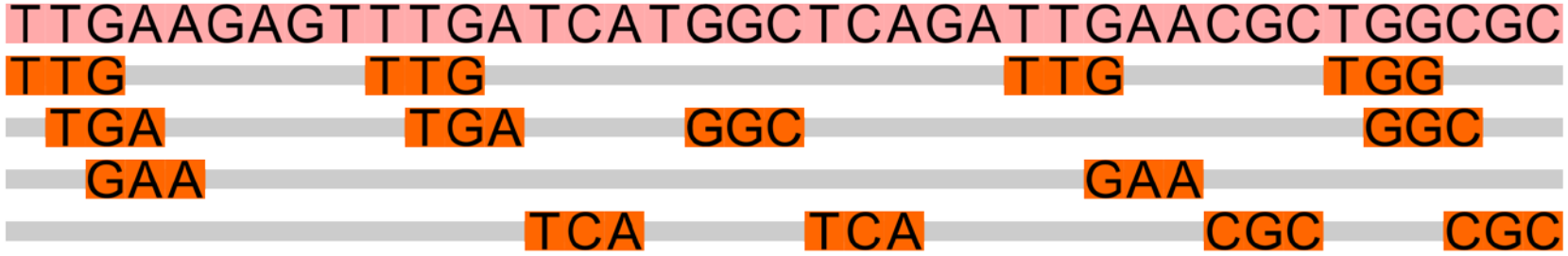
Identify sequences of shared *k*-mers



Genome assembly

De Bruijn graph assembly

Problem #1 – *k*-mers are short?



Genome assembly

De Bruijn graph assembly

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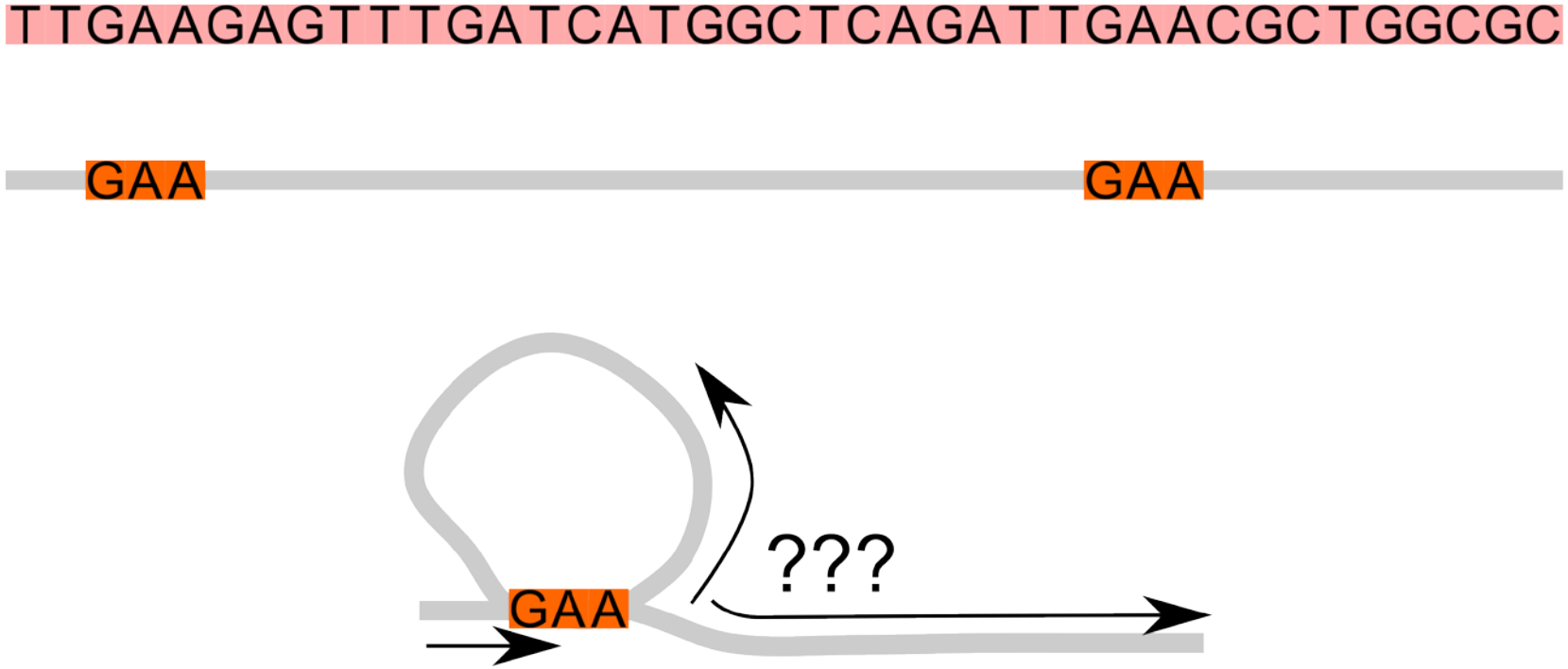
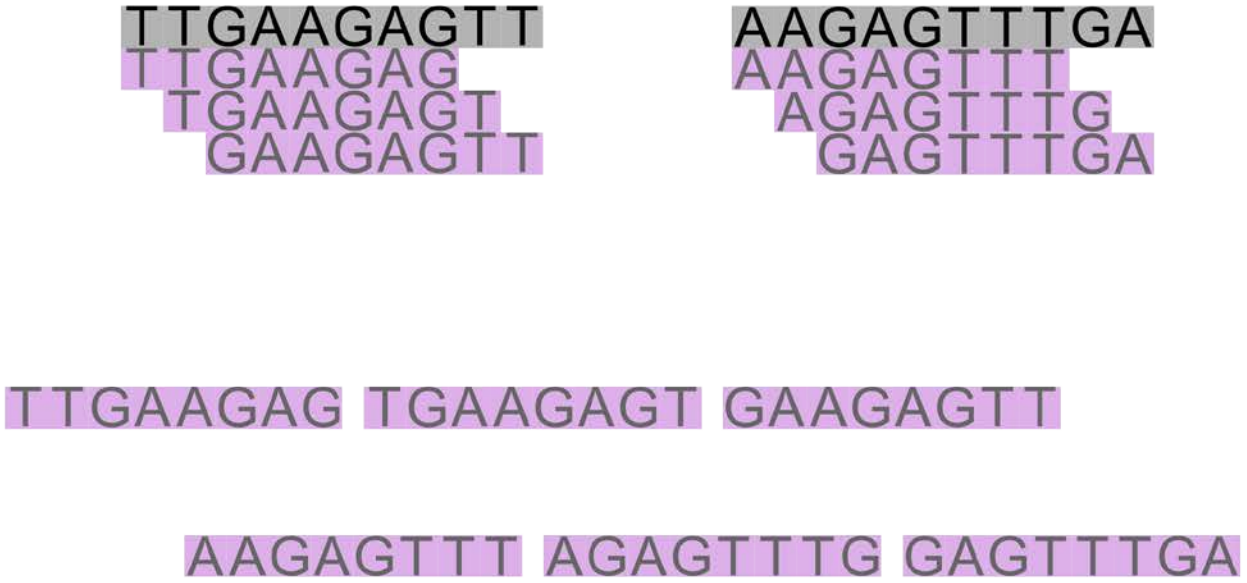


Genome assembly

De Bruijn graph assembly

Problem #2 – *k*-mers are long?



De Bruijn graph assembly

We want a range of *k*-mer sizes

- Short *k*-mers yield higher coverage
- Long *k*-mers assemble longer contigs (jump repeat regions)

Other considerations for picking *k*-mer sizes

- Size cannot be longer than read length
- Always pick odd *k*-mer sizes
- The more sizes you use, the longer assembly will take

K-mers	N. contigs	Longest contig	N50 >2kbp	L50 >2kbp
21, 33, 55	4,239,806	660,812	6,782	12,906
43, 55, 77, 99, 121	2,519,669	1,022,083	7,990	12,673
21, 43, 55, 77, 99, 121	3,388,682	1,022,083	7,789	13,327

Which assembler is best?

There are three good options

- SPAdes
- IDBA-UD
- MegaHIT

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- SPAdes
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In conclusion, it can be said that the choice of assembler should depend on the data at hand and on the exact research question asked. Generally, the best assembly is performed by multi k-mer assemblers such as metaSPAdes, Megahit and IDBA-UD. If micro diversity is not a major issue, and the primary research goal is to bin and reconstruct representative bacterial genomes from a given environment, metaSPAdes should clearly be the assembler of choice. This assembler yields the best contig size statistics while capturing a high degree of community diversity, even at high complexity and low read coverage. If mico diversity is however an issue, or if the degree of captured diversity is far more important than contig lengths, then IDBA-UD or Megahit should be preferred.

Vollmers *et al.* 2017 (https://doi.org/ 10.1371/journal.pone.0169662)

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What are some key considerations?

Biological

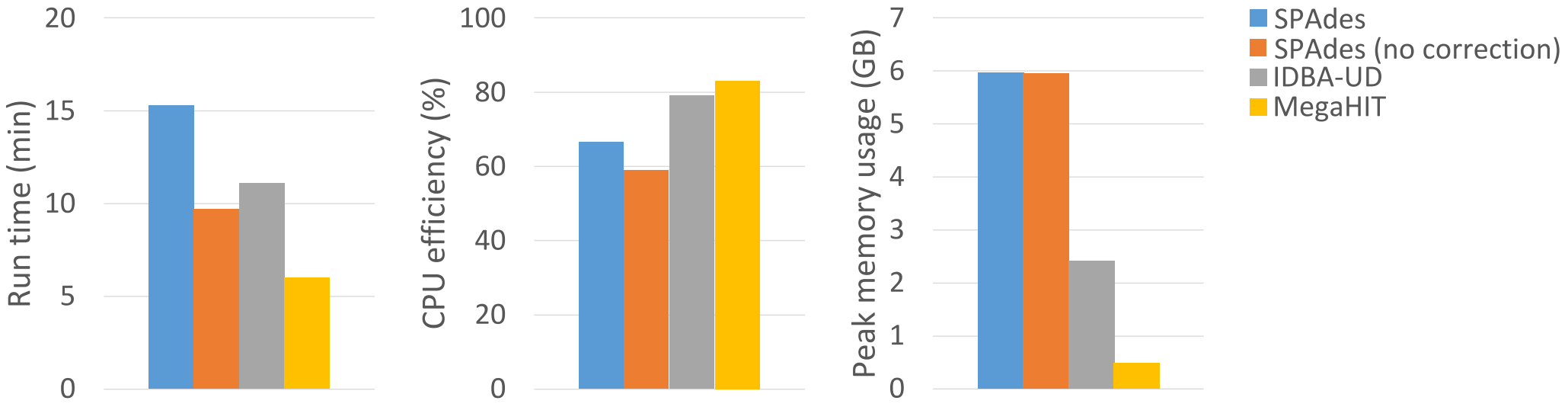
1. What is your hypothesis?
2. What do you want from the data?

Computational and resource

1. How much data do you have?
2. What are your computational resources?
3. What are your ***time*** resources?

Genome assembly

What are some key considerations?



Too much data?

- Consider testing sub-samples when coverage is very high, e.g. 100s or 1000s
- Example: abundant groundwater genome at 2000x coverage in full dataset
- Empirical testing of subsample sizes identified assembly sweet spot

~200x coverage

~2000x coverage

Task: Assembly

Preparing data for assembly

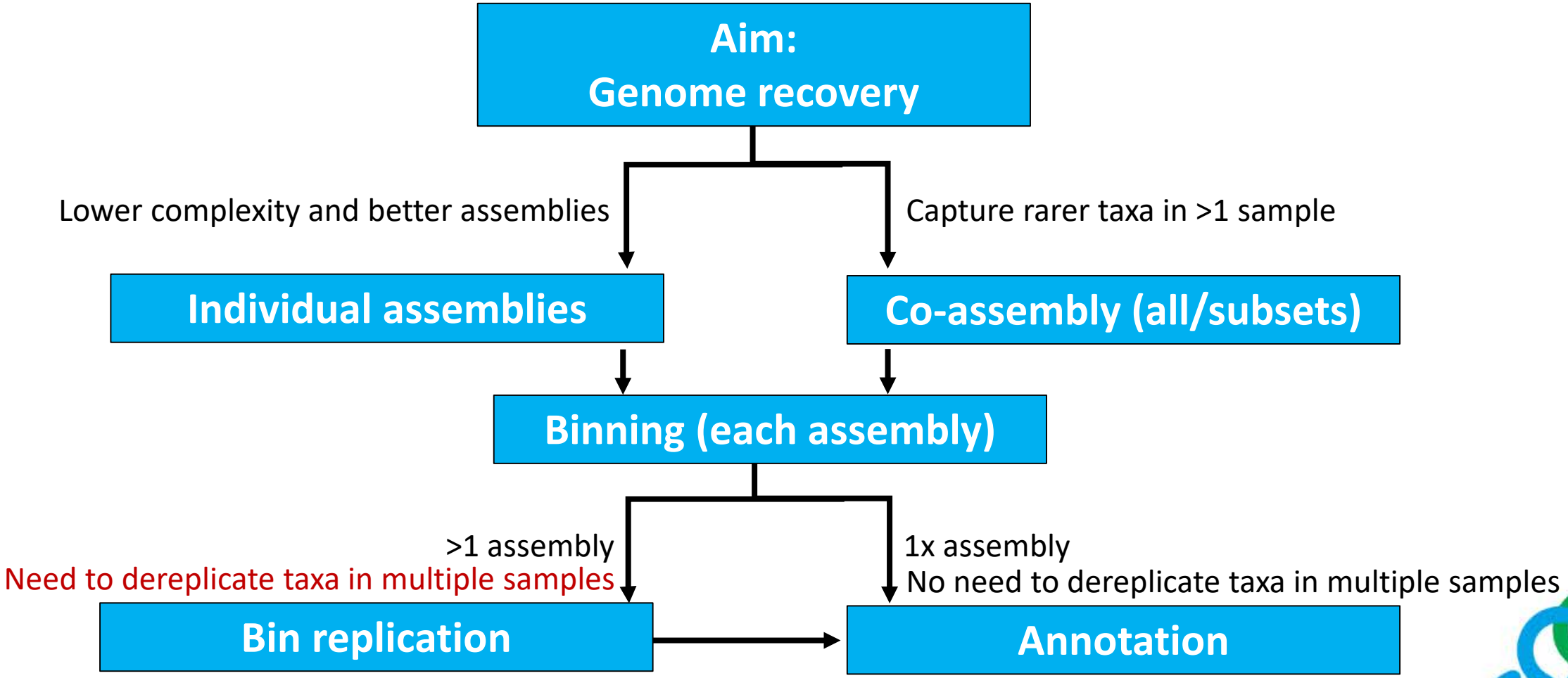
1. Learn to prepare input files for SPAdes and IDBA-UD
2. Configure the basic parameters for assembly

Perform assembly (assemblies)

1. Prepare an assembly job to run under slurm
2. Submit several jobs with varying parameters

Future considerations and Assembly evaluation

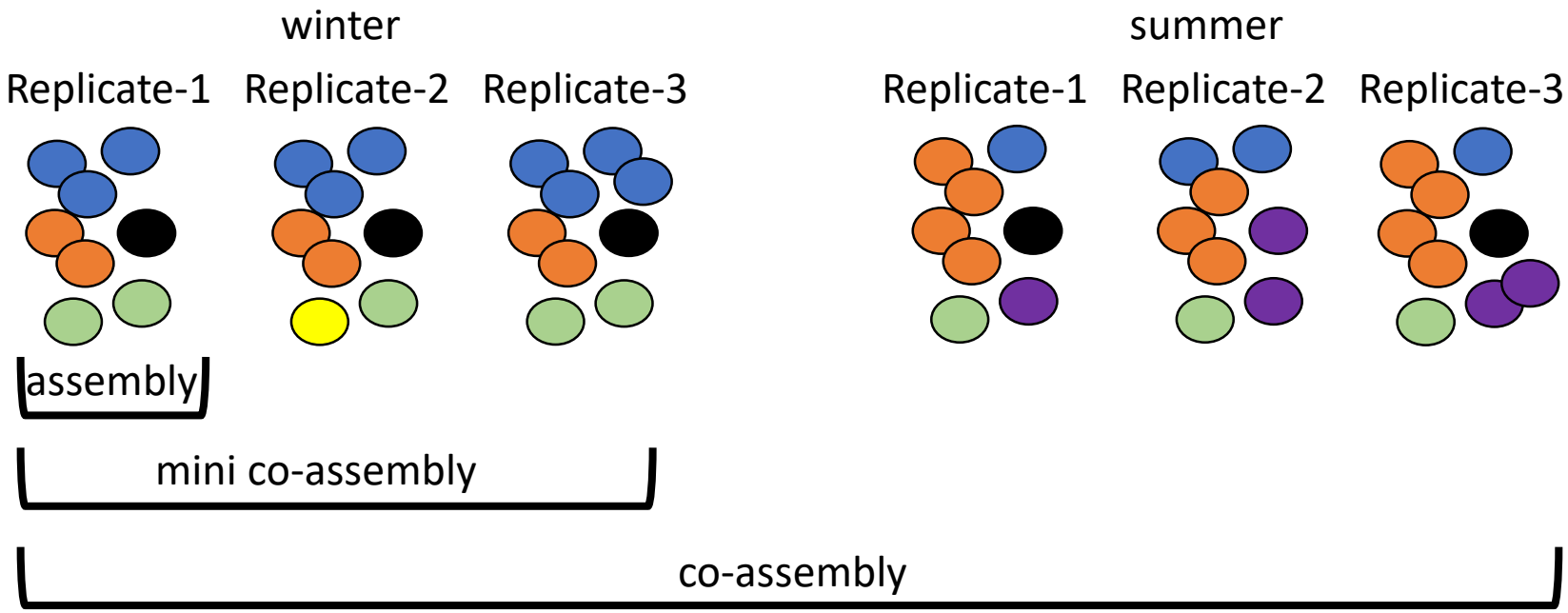
Future considerations



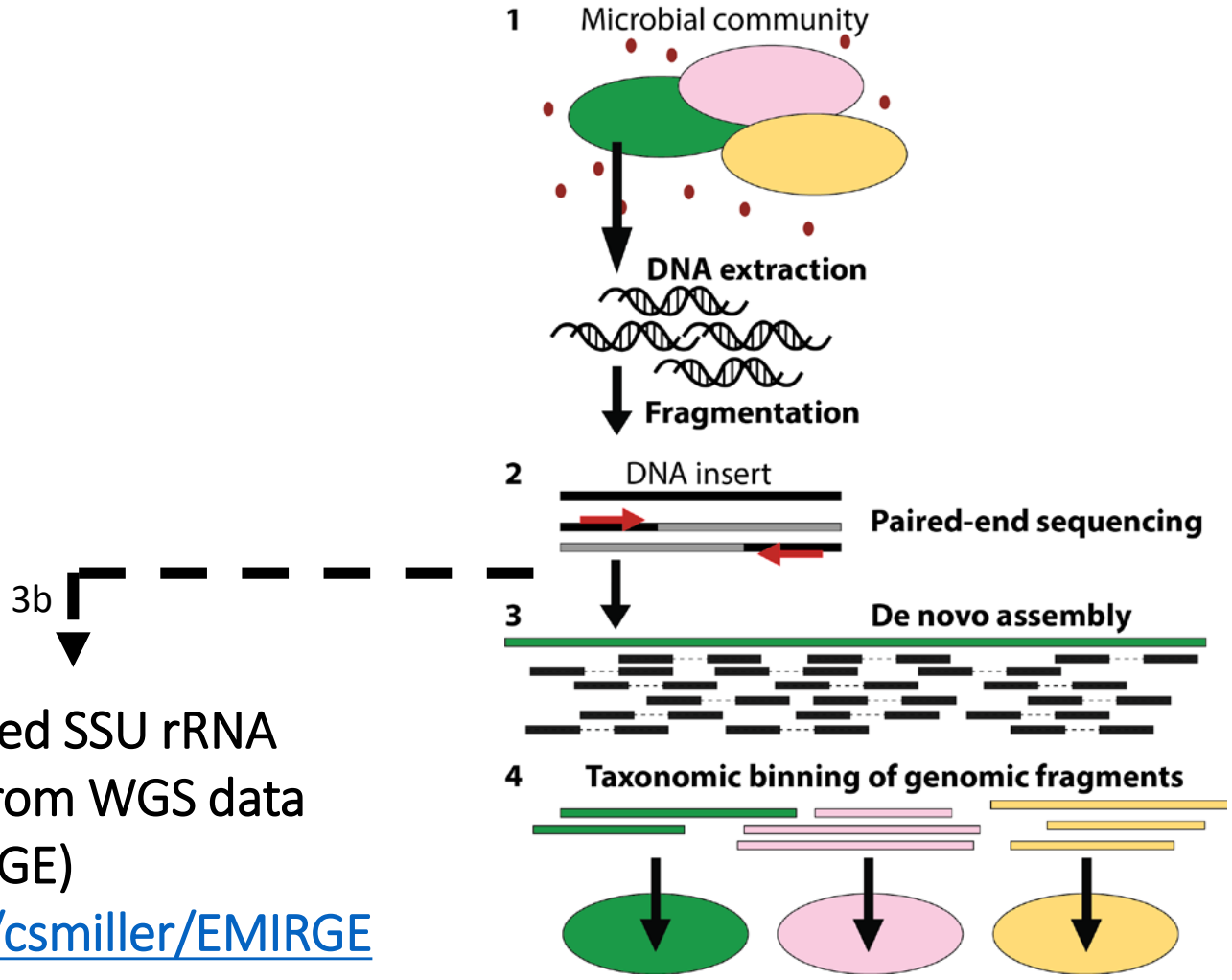
Future considerations

Assembly options:

- **Assembly each community separately**
- **Combine reads and assembly all together (co-assembly)**
- **Combine only reads from the same season (mini co-assemblies)**

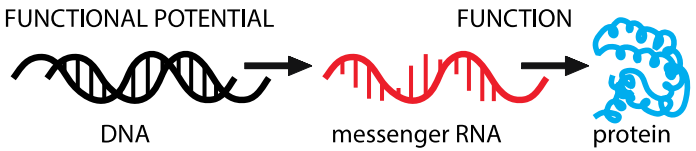


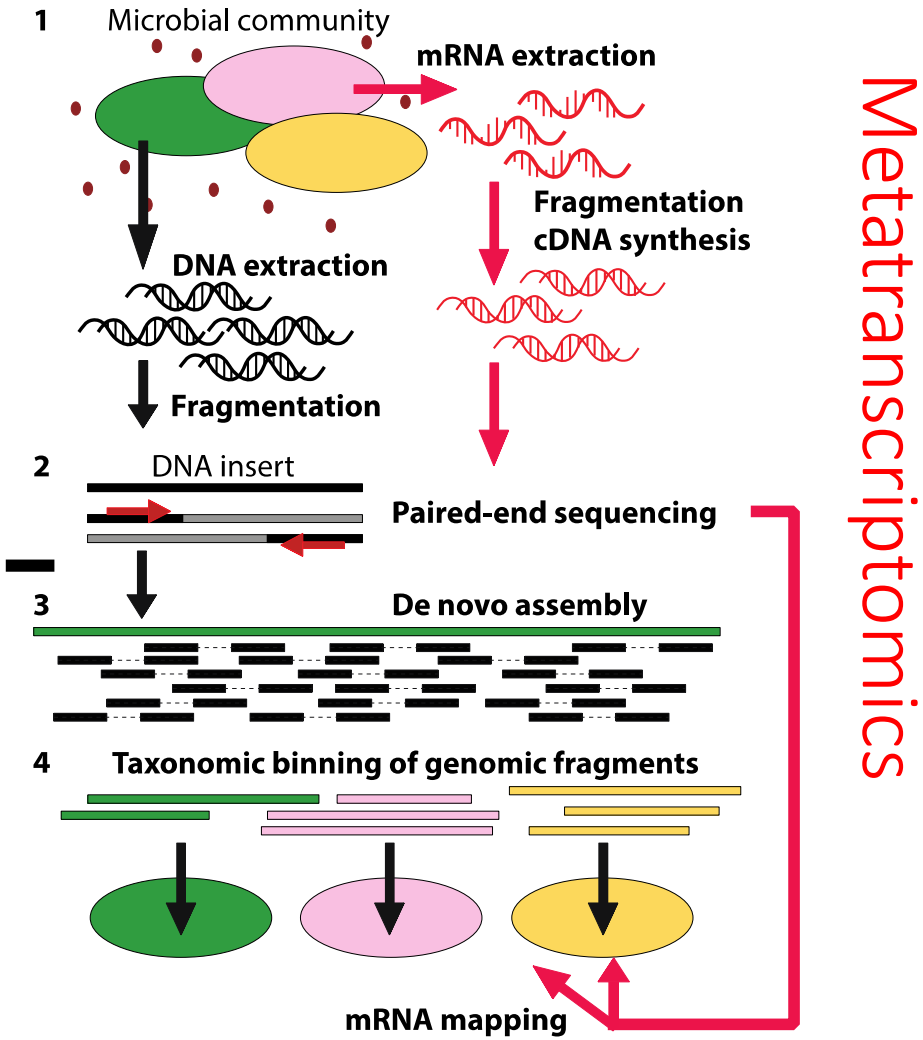
Future considerations: rRNA genes



Reference-guided SSU rRNA reconstruction from WGS data (EMIRGE)
https://github.com/csmiller/EMIRGE

Future considerations: mRNA





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Assembly evaluation

Parameters to use in evaluation:

- **Total length of contigs (= amount assembled)**
- **Total length of contigs usable (e.g. >1,000 bp)**
- **Number of contigs (less is more)**
- **N50 (minimum contig length at 50% of the total genome length)**
- **Length distribution of contigs**
- **Recovery of particular genomes (determined at later stage)**

Task: Assembly evaluation

Go to Github MSS webpage

Task:

- **Run evaluation tool/script in ex5_evaluating_assemblies**