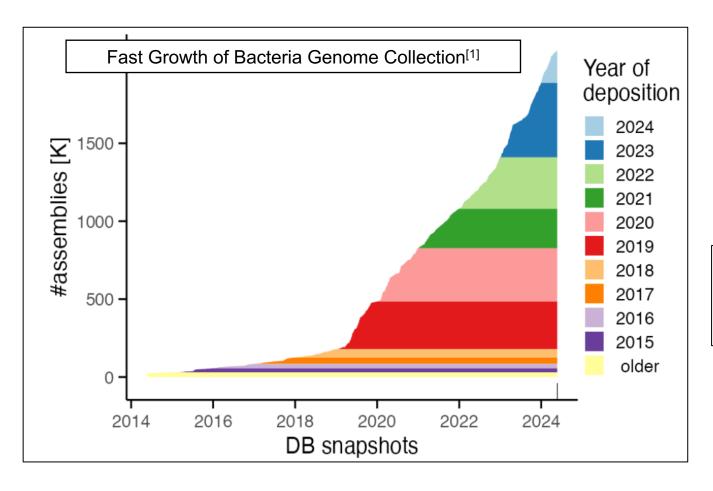
# Optimization For Efficient Compression Of Large Bacterial Genome Collections

### **MOTIVATION:** Larger And Higher Diversity Genome Collections Are Growing Rapidly



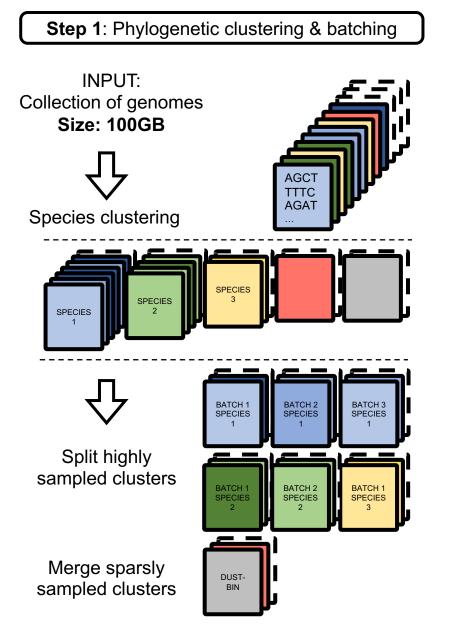
 Large
 Bacterial
 Genome
 Collections:

 661k collection<sup>[2]</sup> (2021)
 n = 661,405

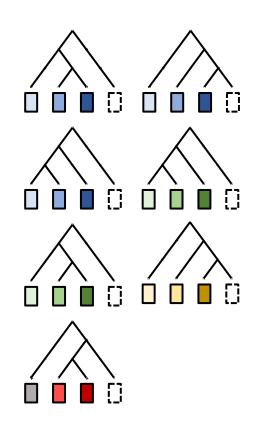
 AllTheBacteria<sup>[3]</sup> (2024)
 n = 2,440,377

 Future
 n >  $10^7$ 

## Phylogenetic Compression<sup>[1]</sup> Improves Compressibility Via Reordering According To The Evolutionary History

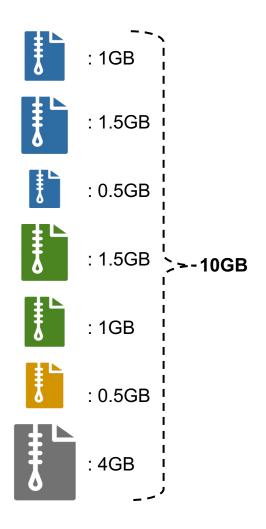


Step 2: Phylogenetic reordering

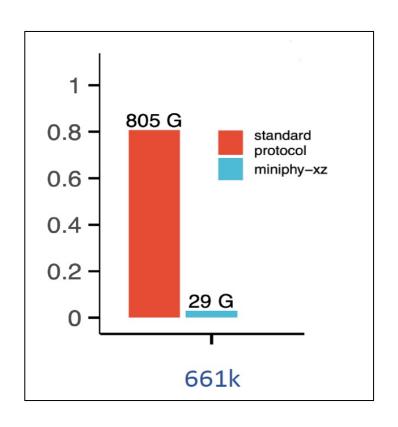


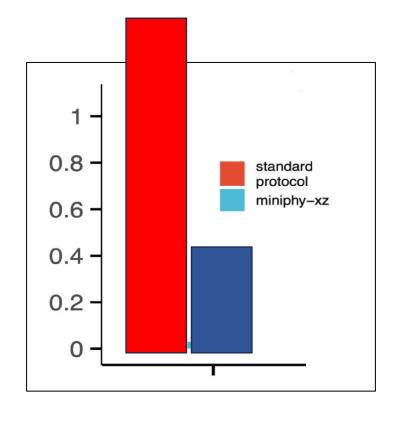
Infer evolutionary trees

Resulting compression



### This Strategy Allows Lossless Compression Of 1-3 Orders Of Magnitude Across Different Genome Collections

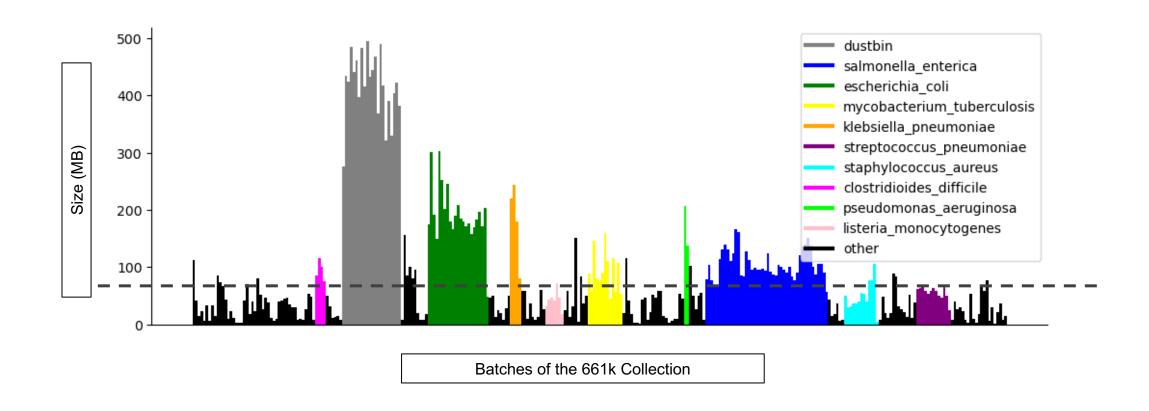




AllTheBacteria - TBA

**RESULTING COMPRESSION** 

### Current Limitation: Batching Results In Non-uniform Post-compression Sizes



# **Current Limitation:** Batching Results In Non-uniform Post-compression Sizes

AllTheBacteria Batches here

Batches of the 661k Collection

# **Consequences:** Negative Impact on Downstream Analysis

**Unbalanced Workloads** 

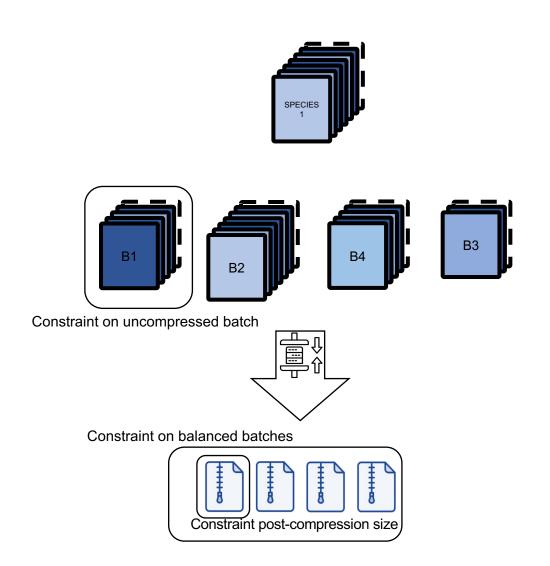
**Hinder Parallelization** 

**Inconsistent Query Times** 

**Memory Overuse** 

**Inefficient Transmission** 

### Challenge: Find Develop An Optimized Batching Strategy For Various Use Cases



### **Batching Problem:**

Given a set of genomes.

Partition it into a set of batches.

User-input parameters for each batch i.e. number of genomes N, uncompressed size U, post-compression size C.

<u>Maximize</u> the compression ratio of the set and in such a way that some <u>constraints</u> are satisfied.

#### **OBJECTIVE:**

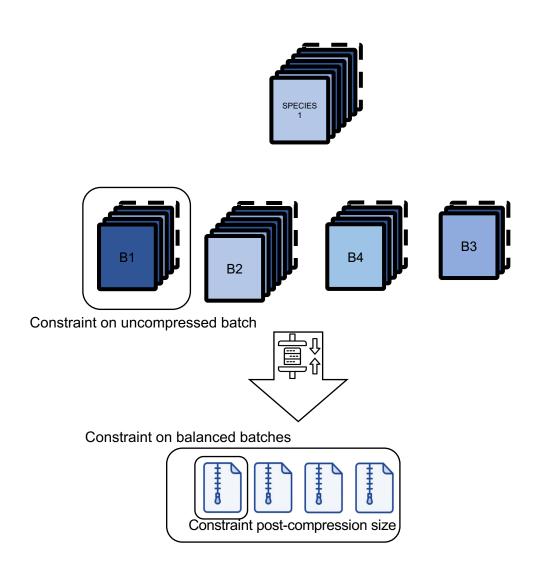
 $min \sum_{i}^{Batches} PostCompressionSize(b_i)$ 

Subjects to:

For for all batches:

 $\begin{aligned} & Cardinality(b_i) \leq N \\ & UncompressedSize(b_i) \leq U \\ & PostCompressionSize(b_i) \leq C \\ & PostCompressionSize(b_i) - PostCompressionSize(b_j) \leq \varepsilon \end{aligned}$ 

### Challenge: Find Develop An Optimized Batching Strategy For Various Use Cases



### **Batching Problem:**

Given a set of genomes.

Partition it into a set of batches.

User-input parameters for each batch i.e. number of genomes N, uncompressed size U, post-compression size C.

<u>Maximize</u> the compression ratio of the set and in such a way that some <u>constraints</u> are satisfied.

#### **OBJECTIVE:**

min  $\sum_{i}^{Batches}$  PostCompressionSize(b<sub>i</sub>)

Subjects to:

For for all batches:

 $\begin{aligned} & Cardinality(b_i) \leq N \\ & UncompressedSize(b_i) \leq U \end{aligned}$ 

PostCompressionSize( $\mathbf{b_i}$ )  $\leq \mathbf{C}$ 

PostCompressionSize( $\mathbf{b_l}$ ) - PostCompressionSize( $\mathbf{b_l}$ )  $\leq \varepsilon$ 

### **Batching of Genomes Collection:** A Familiar yet Novel Problem

### Assumption:

For simplicity, we stop considering genome compression





#### **OBJECTIVE:**

 $min \sum_{i}^{Batches} b_i$ 

Subjects to: For for all batches:

 $\begin{aligned} & \textbf{Cardinality}(\textbf{b_i}) \leq \textbf{N} \\ & \textbf{UncompressedSize}(\textbf{b_i}) \leq \textbf{U} \\ & \textbf{PostCompressionSize}(\textbf{b_i}) \leq \textbf{C} \\ & \textbf{PostCompressionSize}(\textbf{b_i}) - \textbf{PostCompressionSize}(\textbf{b_i}) \leq \boldsymbol{\varepsilon} \end{aligned}$ 

Subjects to: For for all batches:

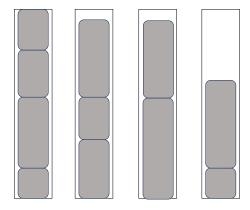
 $\begin{aligned} & Cardinality(b_i) \leq N \\ & UncompressedSize(b_i) \leq U \end{aligned}$ 

This becomes an instance of the classic Bin Packing Problem

### Bin Packing Problem Is One Of The First Studied Combinatorial Optimization Problem

### Bin Packing Problem:

Given a list of item i = 1, ..., n, each having a size  $ci \in R+$ , and an integer value CAPACITY, find the minimum number of bin to pack all items in such a way that the sum of the item sizes in one bin is always smaller than CAPACITY.



### The problem is NP-complete

Classical heuristics are ordered-based algorithms Initially, an empty bin is created.

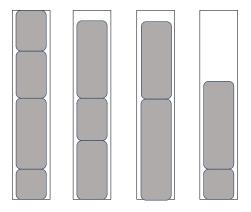
At each step, the next item is selected and packed in a bin. A new bin may be created at each step.

- Next-fit: choose the current bin
- First-fit: choose the first possible bin
- Best-fit: choose largest remaining CAPACITY bin
- Worst-fit: choose smallest remaining CAPACITY bin

### Bin Packing Problem Is One Of The First Studied Combinatorial Optimization Problem

### Bin Packing Problem:

Given a list of item i = 1, ..., n, each having a size  $ci \in R+$ , and an integer value CAPACITY, find the minimum number of bin to pack all items in such a way that the sum of the item sizes in one bin is always smaller than CAPACITY.



### The problem is NP-complete

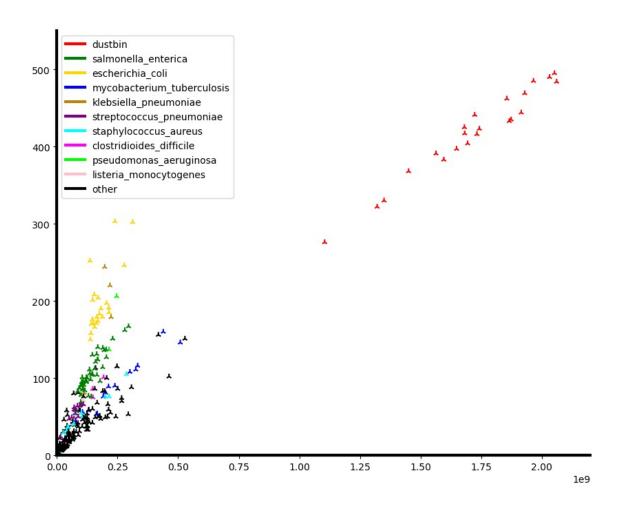
Classical heuristics are ordered-based algorithms Initially, an empty bin is created.

At each step, the next item is selected and packed in a bin. A new bin may be created at each step.

- Next-fit: choose the current bin
- First-fit: choose the first possible bin
- Best-fit: choose largest remaining CAPACITY bin
- Worst-fit: choose smallest remaining CAPACITY bin

Now that we have a strategy to dynamically pack bins based on different parameters, can we estimate the post-compression size without actually compressing the data?

# **Observation**: xz Post-compression 661k Batch Sizes Correlate With Their Distinct Kmers Count



### Ingredient 2: Cardinality estimation using HyperLogLog sketching

Sketches: approximate data structures.

HyperLogLog sketches for cardinality est.: bit patterns,

i.e.  $hash(ATGCG) \rightarrow 00010100$ ,  $hash(CGTAC) \rightarrow 00000010$ .

Fast and efficient UNION operation for sketches.

# HyperLogLog Bin Packing Strategy For Genomes Batching

Pseudocode of Strategy 1

# HyperLogLog Load Balancing Strategy For Genomes Batching

Pseudocode of Strategy 2

# Recap of the Three Batching Strategies

### Comparisons of the batching strategie

