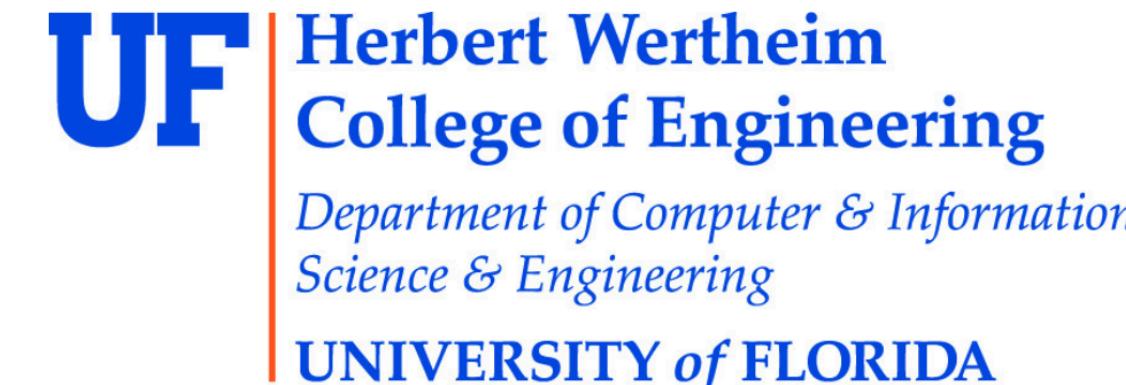


Pan-genomic indexes for robust classification of nanopore and metagenomic reads

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Genome Informatics
November 3-5, 2021



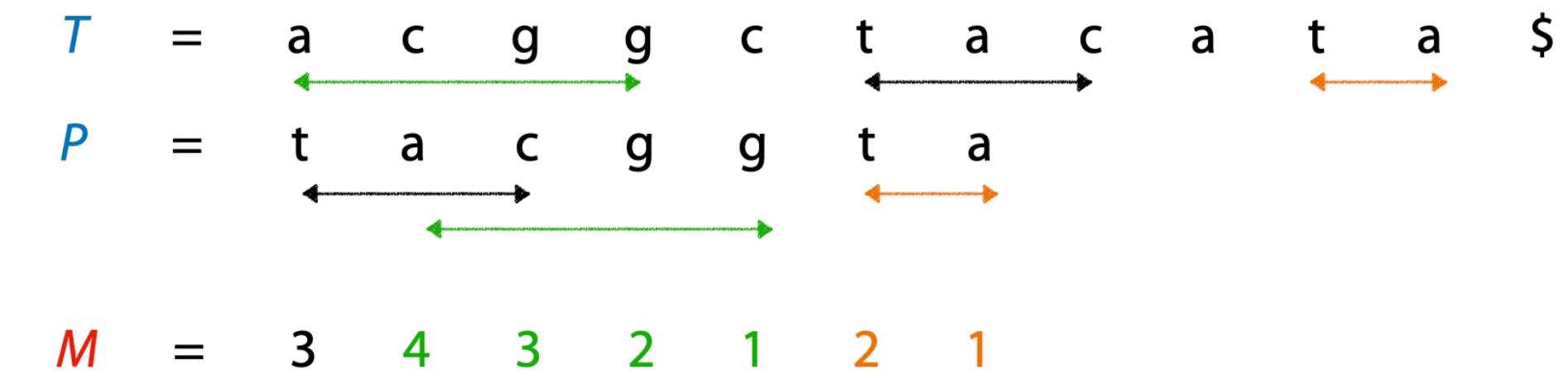
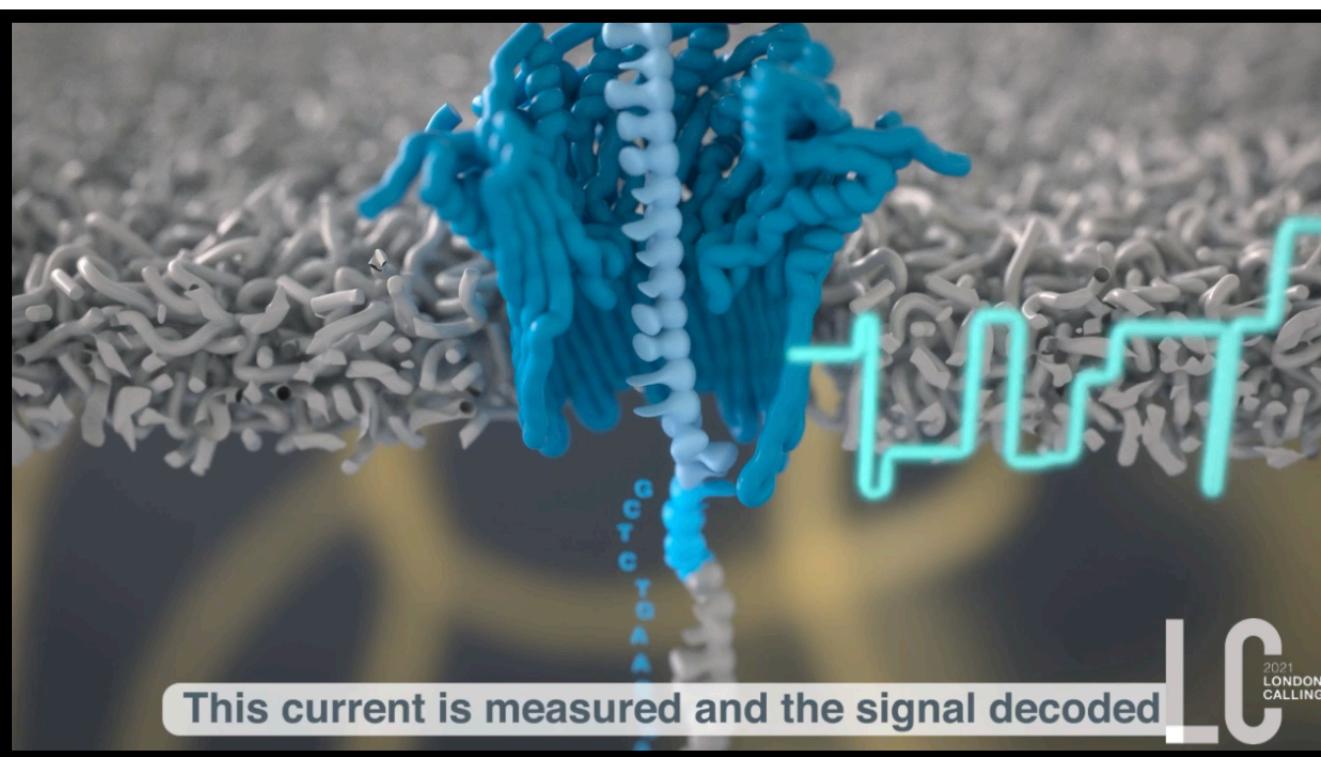
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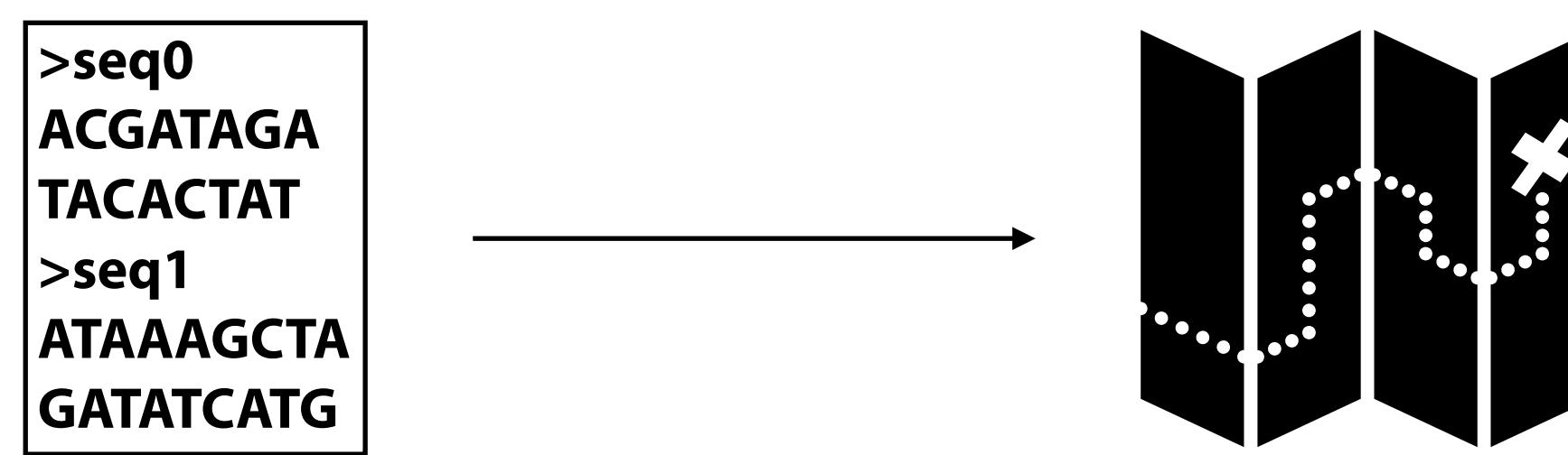
³ Faculty of Computer Science, Dalhousie University, Halifax, NS, CAN

Overview of Presentation

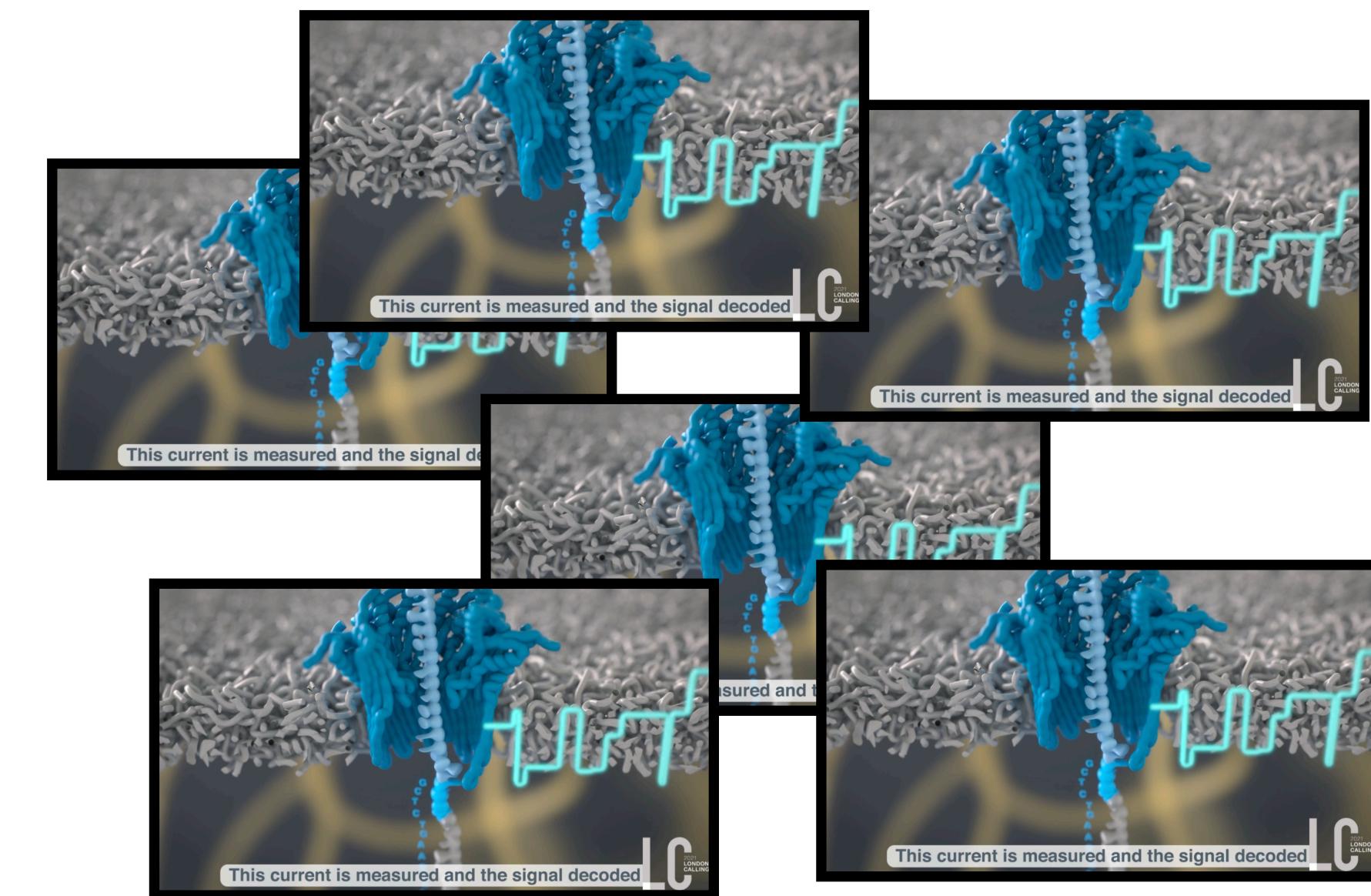
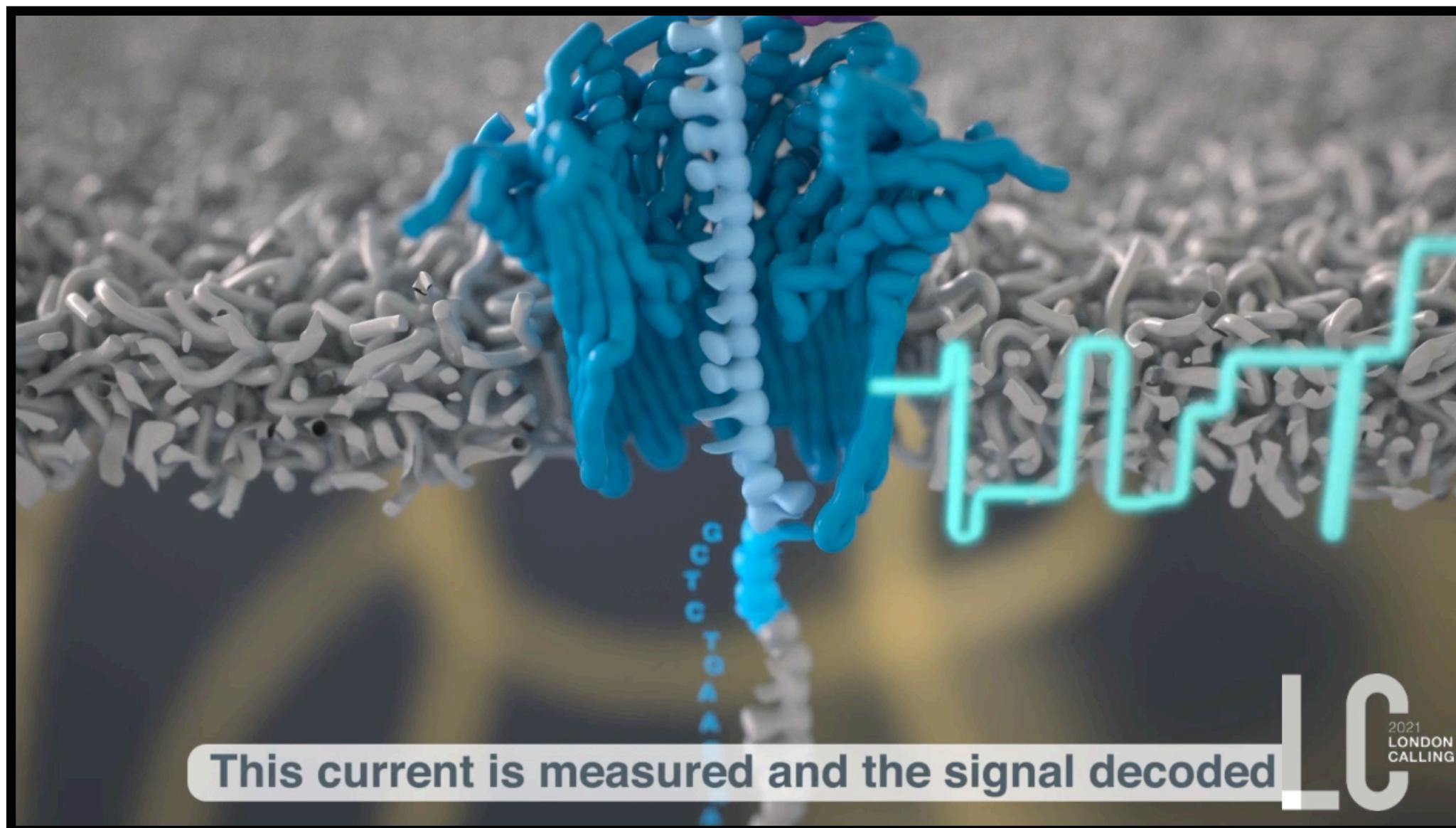
- # ► Development of SPUMONI for classification of nanopore reads



- ## ► Scaling SPUMONI to handle larger pan-genomes more efficiently



Nanopore Sequencing



- ▶ Allows users to perform **targeted sequencing** using software
- ▶ UNCALLED & Readfish allow users to **target sequences** but not optimized for large, and repetitive databases

| Motivation: A need for faster methods to classify reads against large, repetitive databases

Method Overview

Check Max Rossi's Poster 150

SPUMONI - Streaming PseUdo MONI¹

- ▶ Makes rapid targeting decisions based on input database
 - Key Intuition: A read's MS/PMLs with respect to a reference can reveal if there appears to be “good” approximate match to reference
- ▶ Uses the r-index² to enable efficient indexing of large, repetitive collections
 - Number of runs in BWT, r, typically grows sub-linearly w.r.t to length of input sequence, n.
- ▶ Extends MONI¹ in two key areas
 - Adds a “null distribution” and hypothesis testing framework for finding “significant” matches
 - Replaces MONI’s “batch” matching statistic (MS) algorithm with a faster, streaming algorithm
 - ▶ Calculates new quantity called pseudo-matching lengths (PMLs)

¹Rossi, M., Oliva, M., Langmead, B., Gagie, T., & Boucher, C. (2021). MONI: A pangenomics index for finding MEMs. *Proc. RECOMB*.

²Mun, T., Kuhnle, A., Boucher, C., Gagie, T., Langmead, B., and Manzini, G. (2020). Matching reads to many genomes with the r-index. *J. Comput. Biol.* 27, 514–518

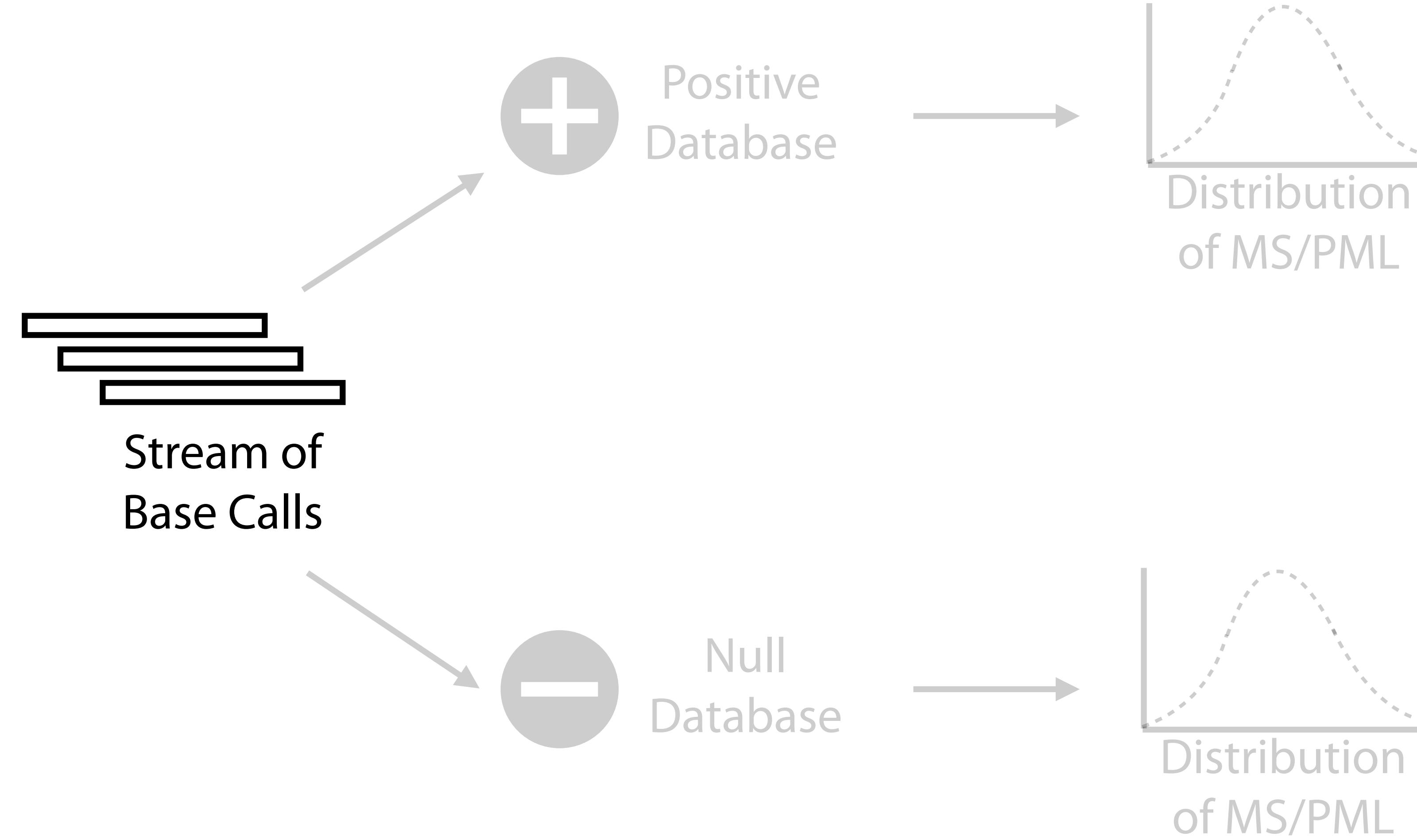
Matching Statistics

- The matching statistics of P w.r.t to T is an array M of length m where $M[i]$ is the length of the longest prefix of the pattern $P[i..m]$ that occurs in text T
 - Let T be a text of length n , and P be a pattern of length m
- Think of matching statistics like they are half MEMs (Maximally Exact Matches)

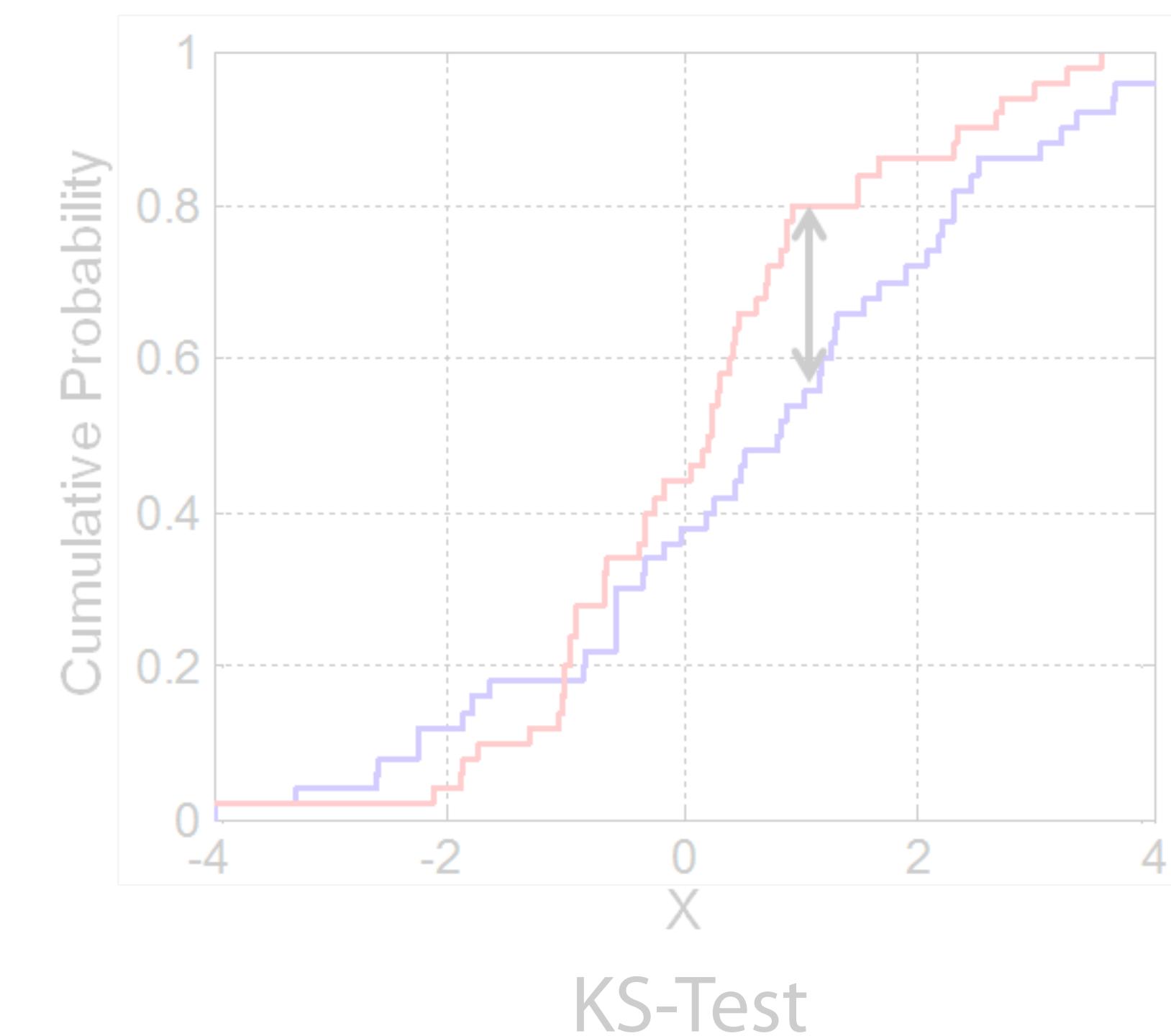
- Example:

Positions:	0	1	2	3	4	5	6	7	8	9	10	11
$T =$	a	c	g	g	c	t	a	c	a	t	a	\$
$P =$	t	a	c	g	g	t	a					
$M =$	3	4	3	2	1	2	1					

SPUMONI Approach



Statistical Tests:



Results - Real Mock Community Experiment

- **Question:** Can using a pan-genome reference allow us to target a particular strain that is not present in the reference? and how does it compare on time and memory?
- Using **real** mock community reads¹ where we want to “target” the **yeast** reads, and eject all the **microbial** species

Reference:	One genome ref (7 genomes from 7 species)		Pan-genome ref (3537 genomes from 7 species)	
Approach:	SPUMONI	minimap2	SPUMONI	minimap2
Accuracy:	86.72	87.82	96.02	97.52

- **SPUMONI is ...**
- 12X faster than minimap2
 - Uses 4X less memory than minimap2

- **Answer:**
- ① Yes, using a pan-genome reference, allowed us to target the ZymoMC strains
 - ② Faster and uses less memory than minimap2 with similar classification metrics

¹Kovaka, S., Fan, Y., Ni, B. et al. Targeted nanopore sequencing by real-time mapping of raw electrical signal with UNCALLED. *Nat Biotechnol* **39**, 431–441 (2021).

Pillars of SPUMONI

- What are the key ideas for why SPUMONI outperforms alignment in classifying reads?

Key Ideas	Why?
SPUMONI is faster .	Non-alignment method
SPUMONI is scalable to large databases .	Use of r-index ¹
SPUMONI's classification is robust .	Non-parametric, Non-kmer method

 Motivated us to **push SPUMONI into other domains** like metagenomic classification!

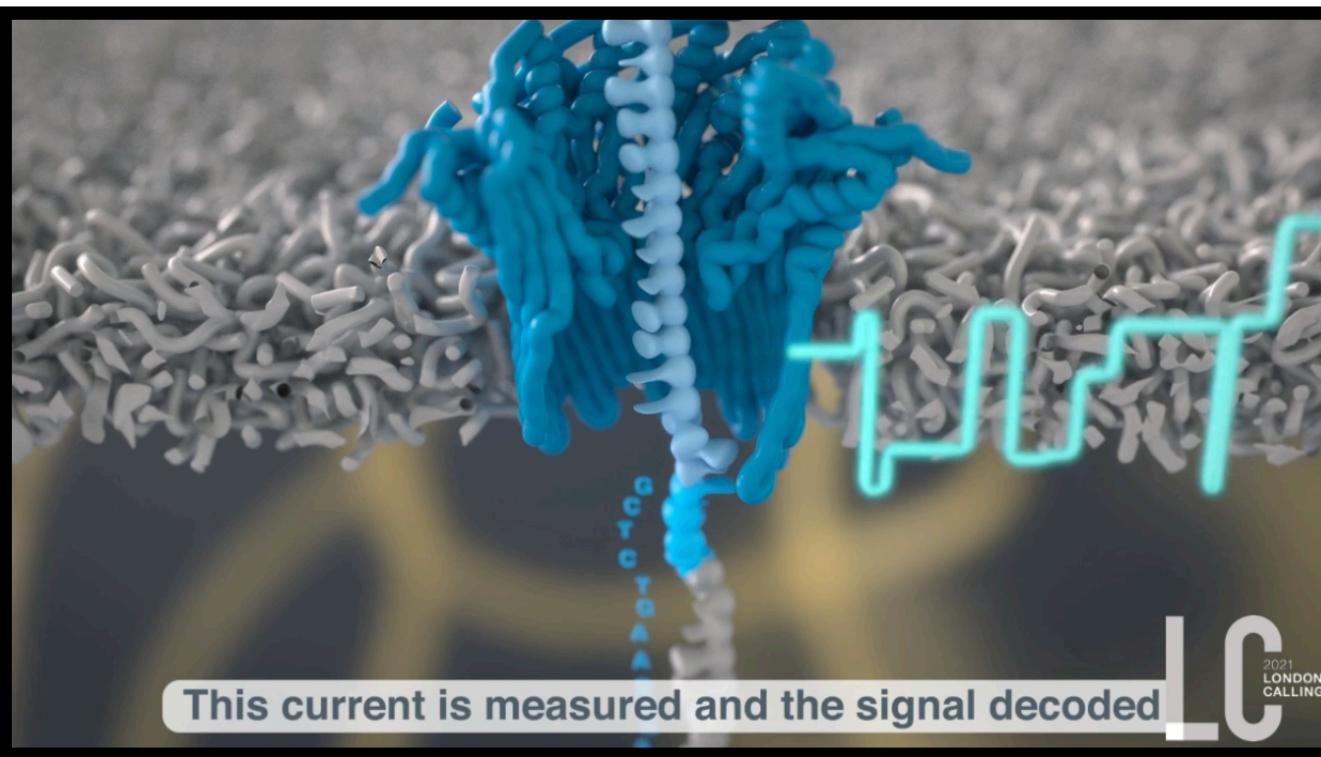
- For details on additional specifics of method and results:

Ahmed, O., Rossi, M., Kovaka, S., Schatz, M. C., Gagie, T., Boucher, C., & Langmead, B. (2021). Pan-genomic Matching Statistics for Targeted Nanopore Sequencing. iScience, 102696.

¹ Mun, T., Kuhnle, A., Boucher, C., Gagie, T., Langmead, B., and Manzini, G. (2020). Matching reads to many genomes with the r-index. J. Comput. Biol. 27, 514–518

Overview of Presentation

- Development of SPUMONI for classification of nanopore reads

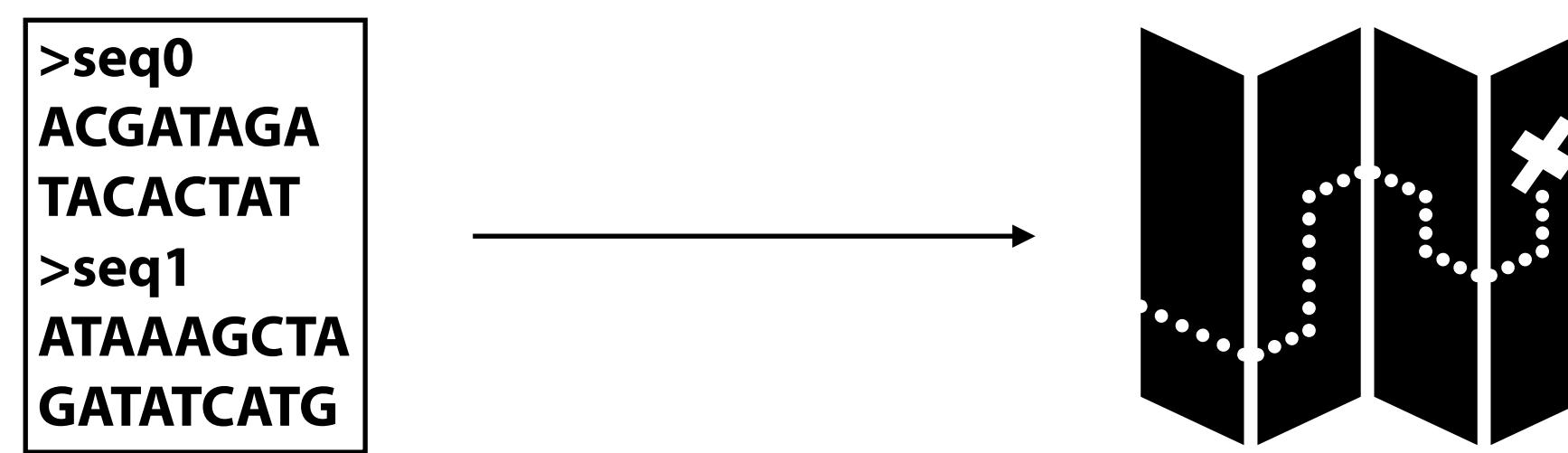


$T = \text{a } \xleftarrow{\text{green}} \text{c } \xrightarrow{\text{green}} \text{g } \xleftarrow{\text{green}} \text{g } \xleftarrow{\text{green}} \text{c } \xleftarrow{\text{black}} \text{t } \xleftarrow{\text{black}} \text{a } \xleftarrow{\text{black}} \text{c } \xleftarrow{\text{black}} \text{a } \xleftarrow{\text{orange}} \text{t } \xleftarrow{\text{orange}} \text{a } \$$

$P = \text{t } \xleftarrow{\text{black}} \text{a } \xleftarrow{\text{black}} \text{c } \xleftarrow{\text{green}} \text{g } \xleftarrow{\text{green}} \text{g } \xleftarrow{\text{black}} \text{t } \xleftarrow{\text{orange}} \text{a }$

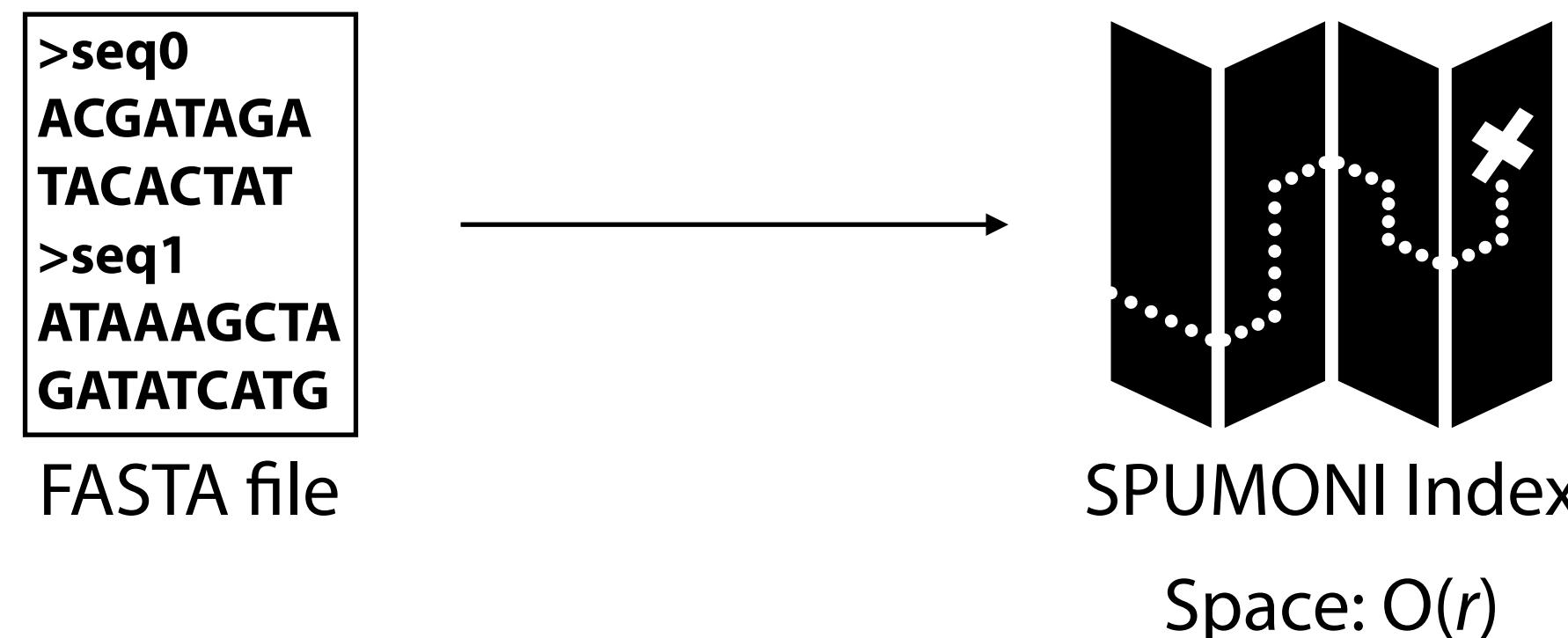
$M = \begin{matrix} 3 & 4 & 3 & 2 & 1 & 2 & 1 \end{matrix}$

- Scaling SPUMONI to handle larger pan-genomes more efficiently



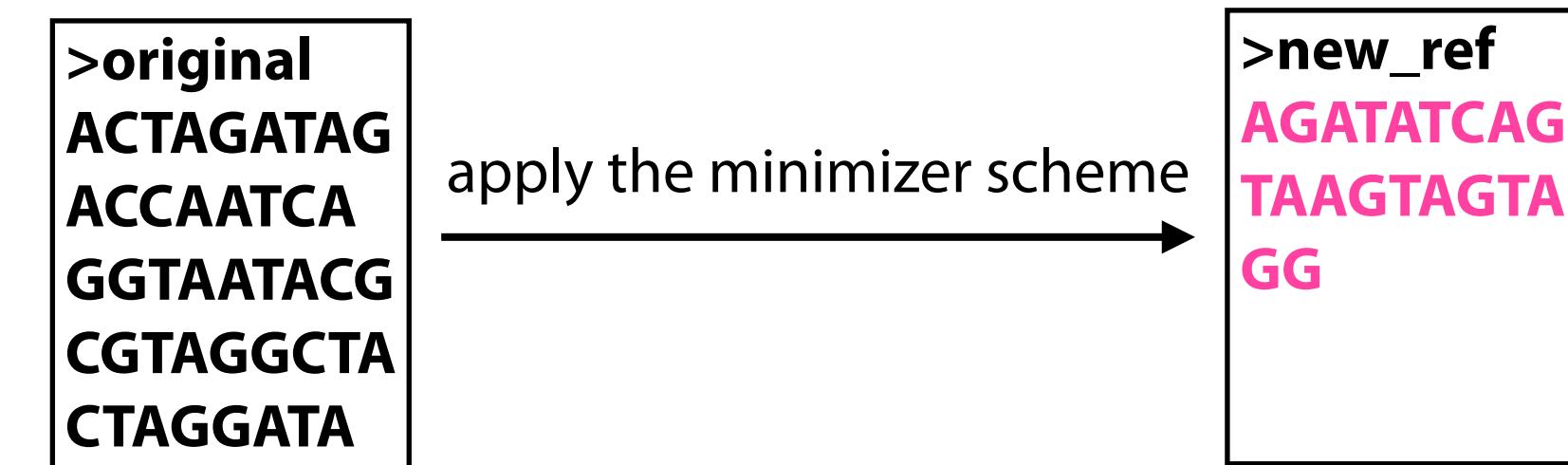
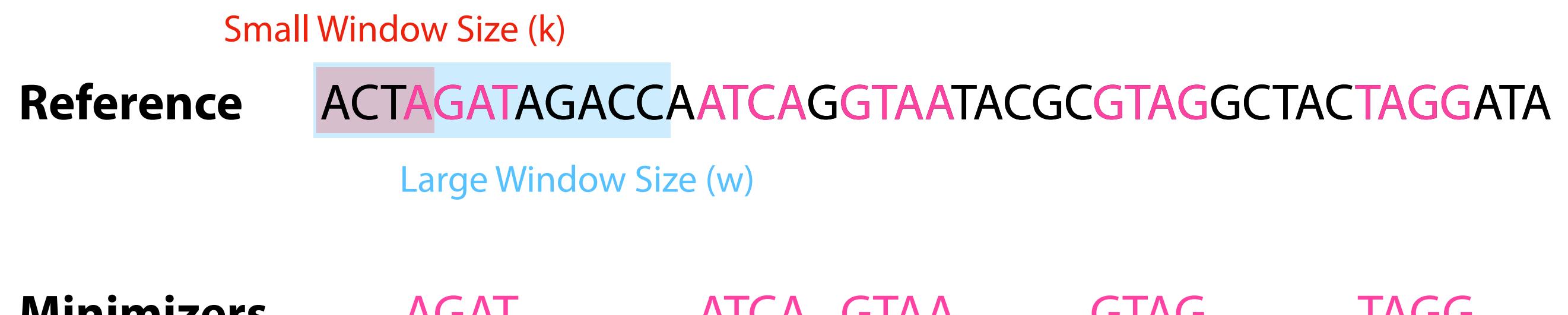
Speeding Up SPUMONI

► How can we speed up SPUMONI?



| Key Idea: Extract and concatenate the minimizers¹ to generate a smaller reference.

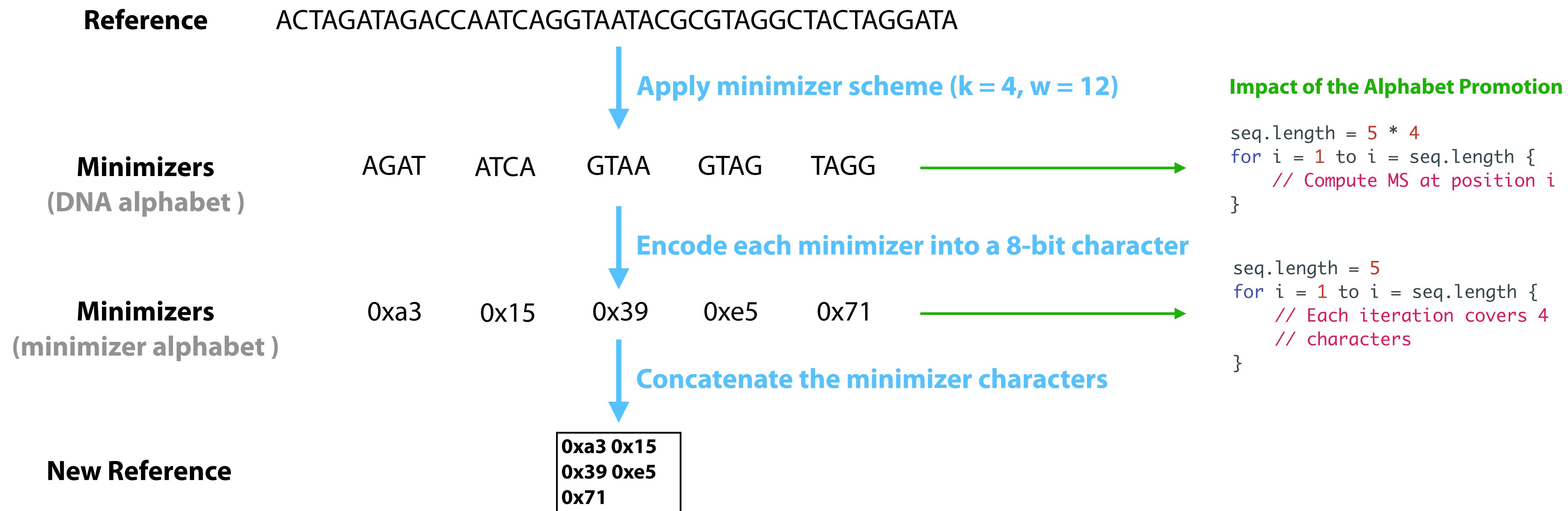
► How did we use the minimizer concept?



| We can apply the same scheme to reads, and thereby index this **smaller** reference.

Speeding Up SPUMONI

- And we can take it even further ...
 - Ekim et al. (2021)¹ use minimizer-alphabet for de Bruijn graphs
- Let's take a look at how we use the minimizer-alphabet in SPUMONI ...

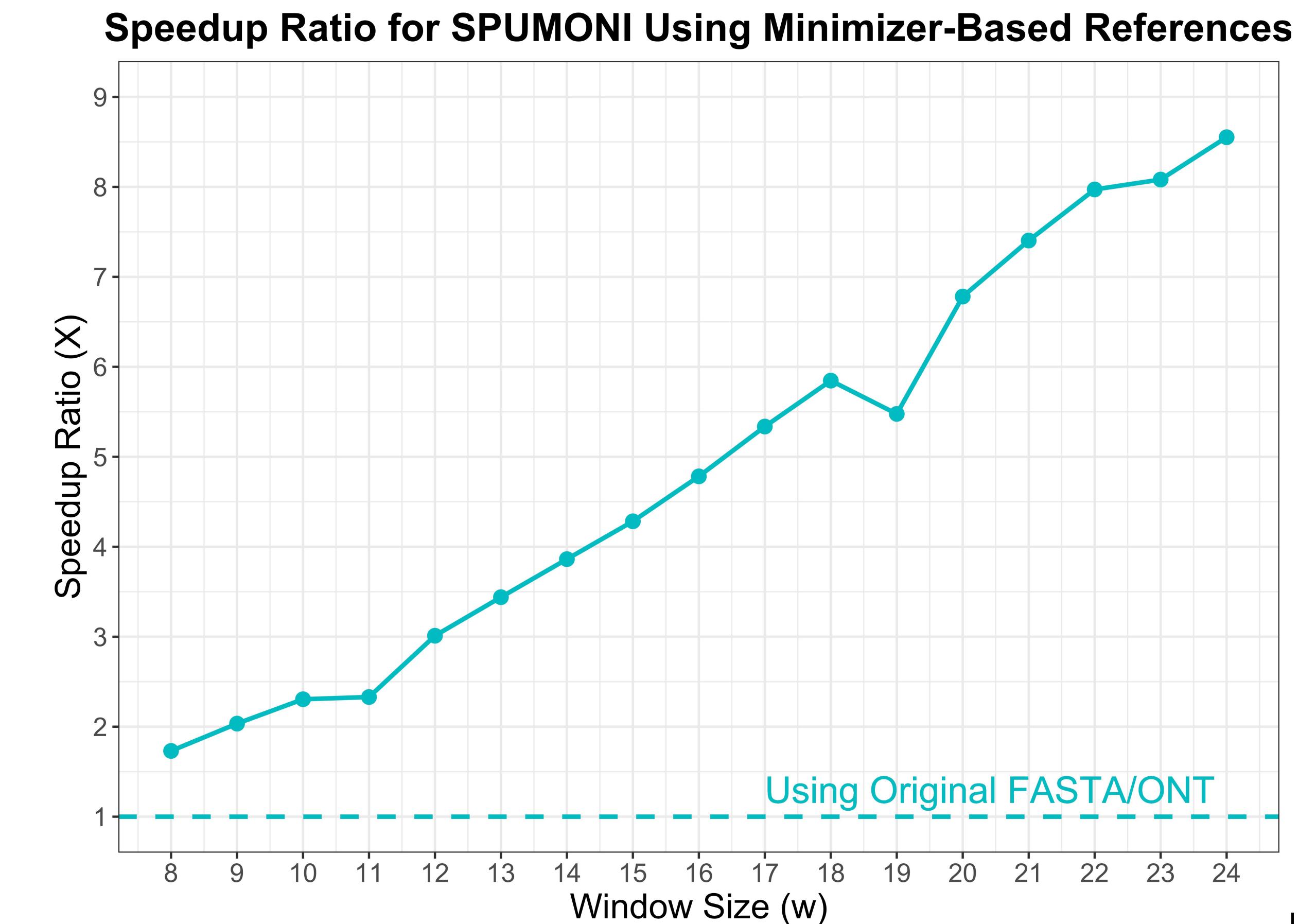
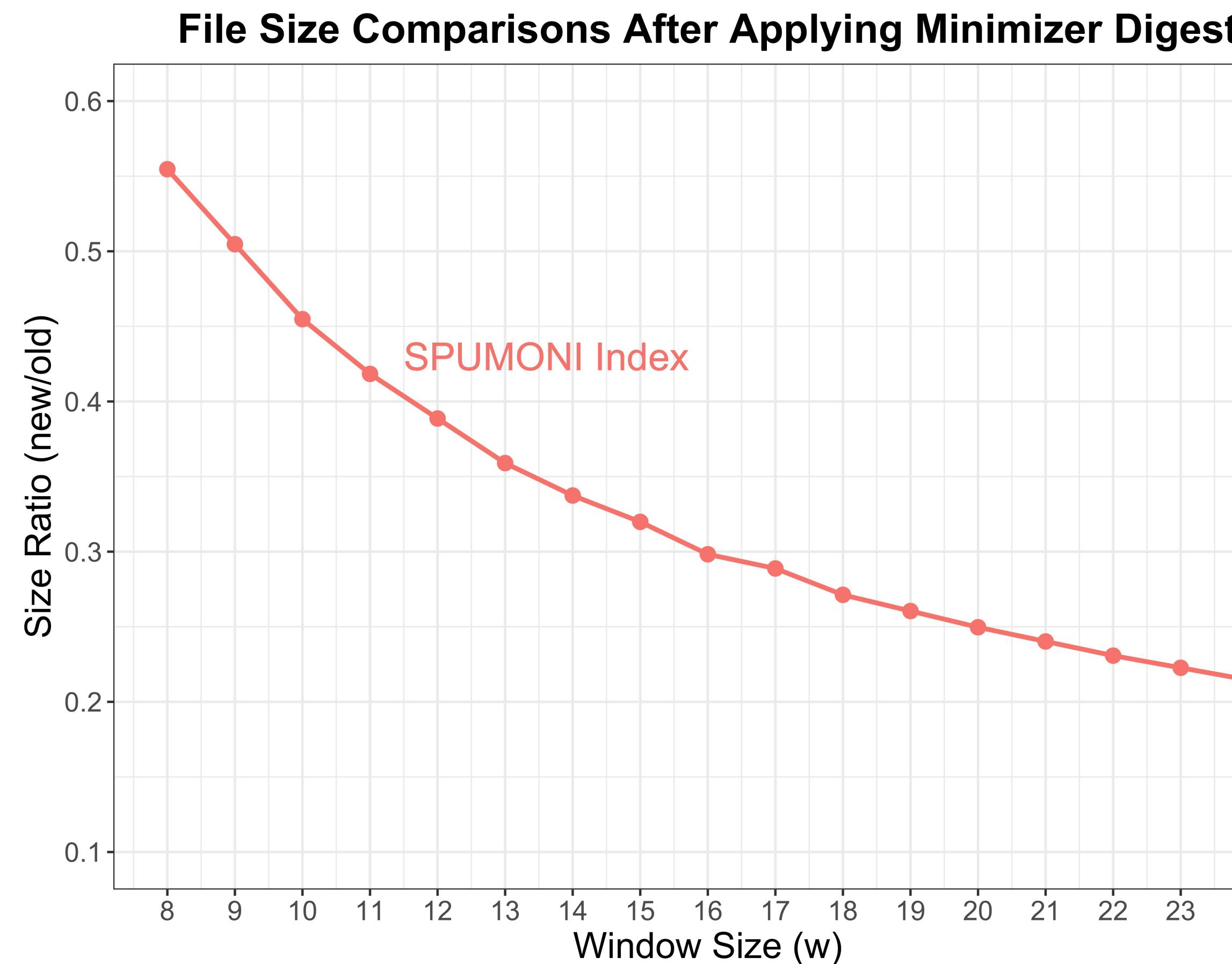


| Combining the minimizer scheme with alphabet promotion leads to even smaller references.

¹ Ekim, B., Berger, B., & Chikhi, R. (2021). Minimizer-space de Bruijn graphs: Whole-genome assembly of long reads in minutes on a personal computer. *Cell Systems*.

Results - Using Minimizer-Based References

- ▶ **Question:** How much smaller will the index be when applying this minimizer scheme where $k = 4$ to the reference?
- ▶ **Answer:** *Built an index over 1833 E. coli genomes¹ (~9 GB) to see ...*

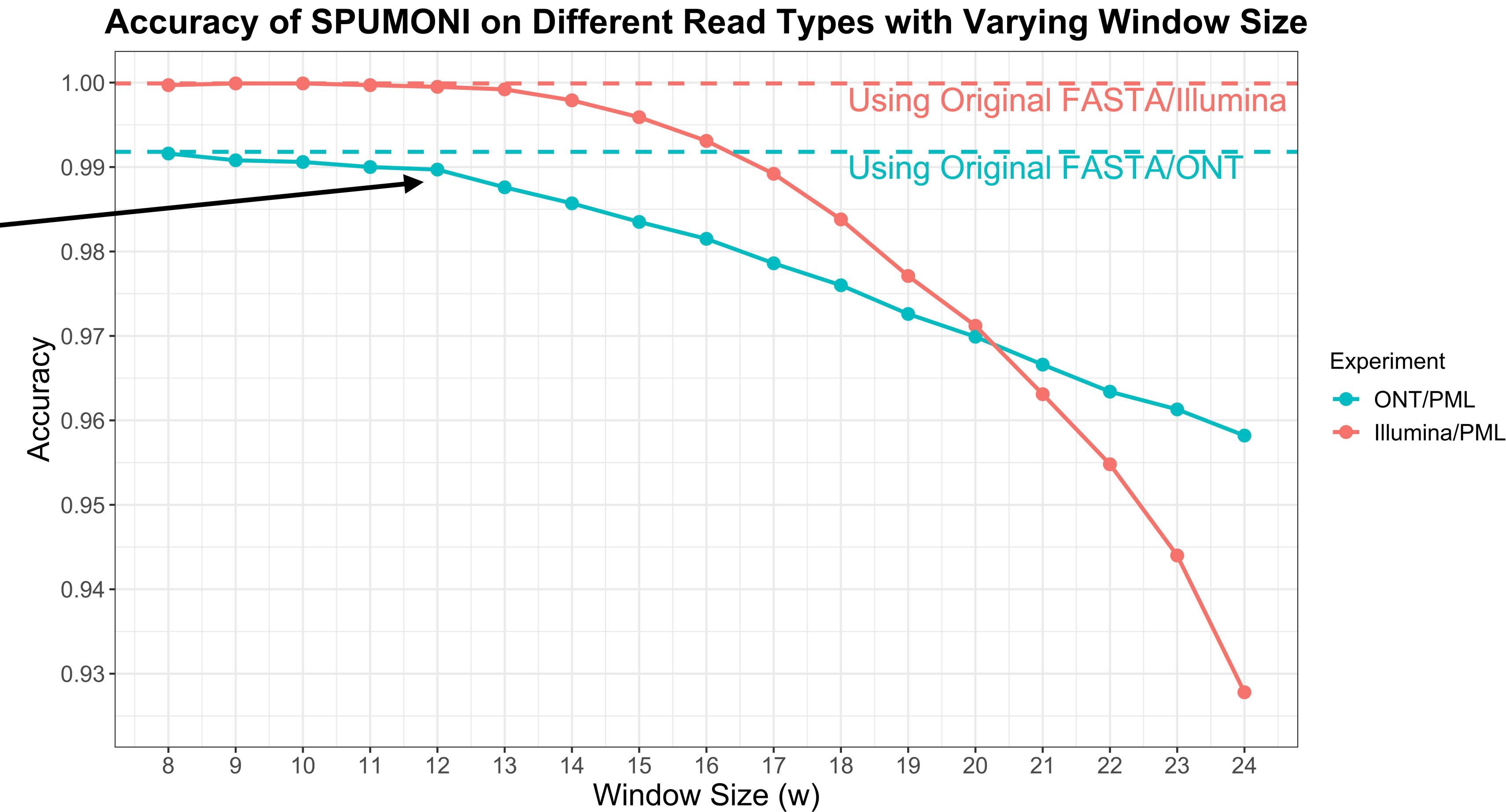


¹O'Leary NA et al. Reference sequence (RefSeq) database at NCBI: current status, taxonomic expansion, and functional annotation. *Nucleic Acids Res.* 2016.

Results - Using Minimizer-Based References

- **Question:** How will this reduction in reference size affect our binary classification ability?
- **Answer:** Simulated Human¹ and E. coli reads², and used our E. coli index to binary classify the reads ...

At w = 12, we are
indexing a file **1/5**
the original!



¹ Nurk S, Koren S, Rhie A, Rautiainen M, et al. The complete sequence of a human genome. bioRxiv, 2021.

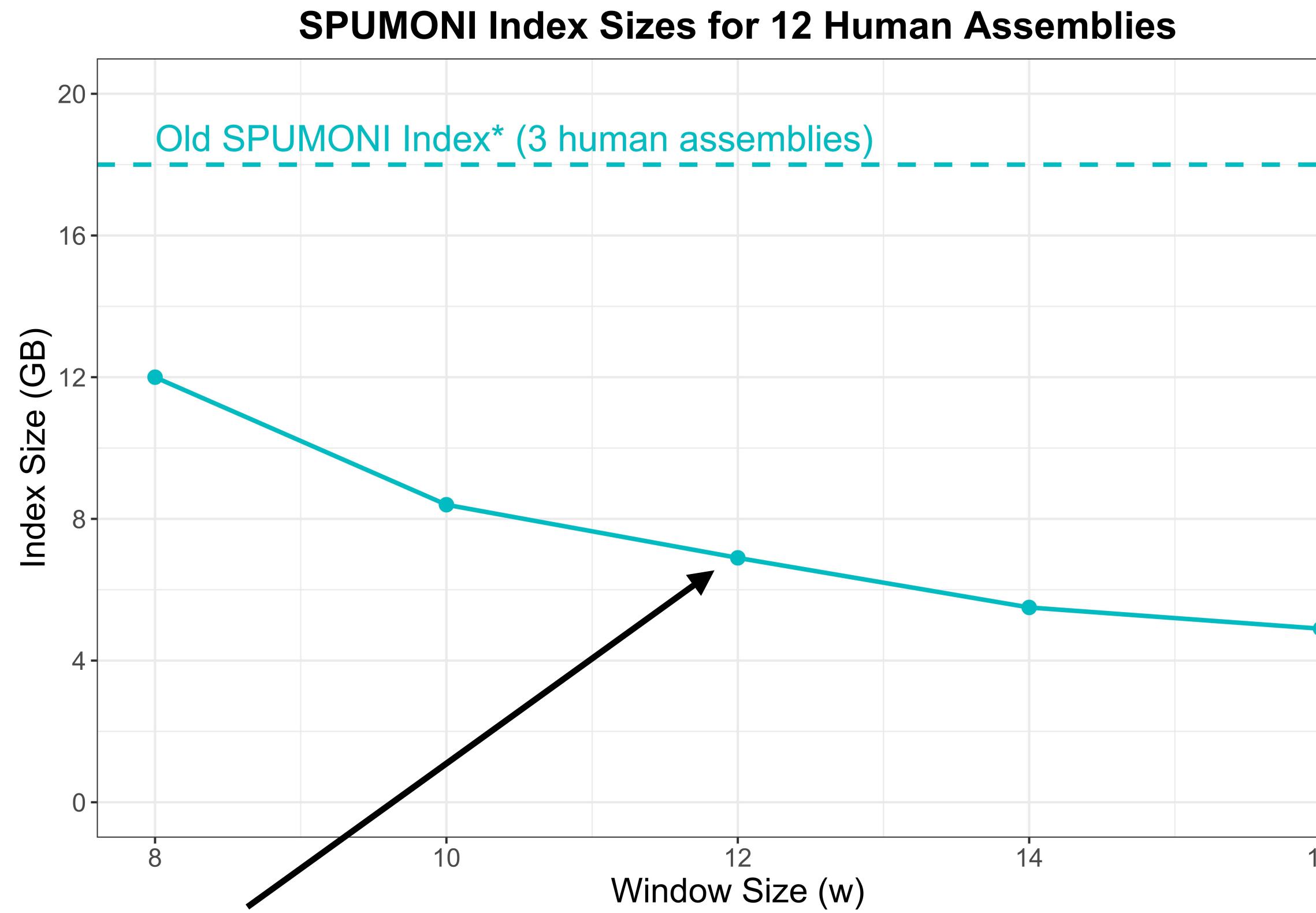
² O'Leary NA, et al. Reference sequence (RefSeq) database at NCBI: current status, taxonomic expansion, and functional annotation. Nucleic Acids Res. 2016.

Results - Indexing 12 Human Assemblies

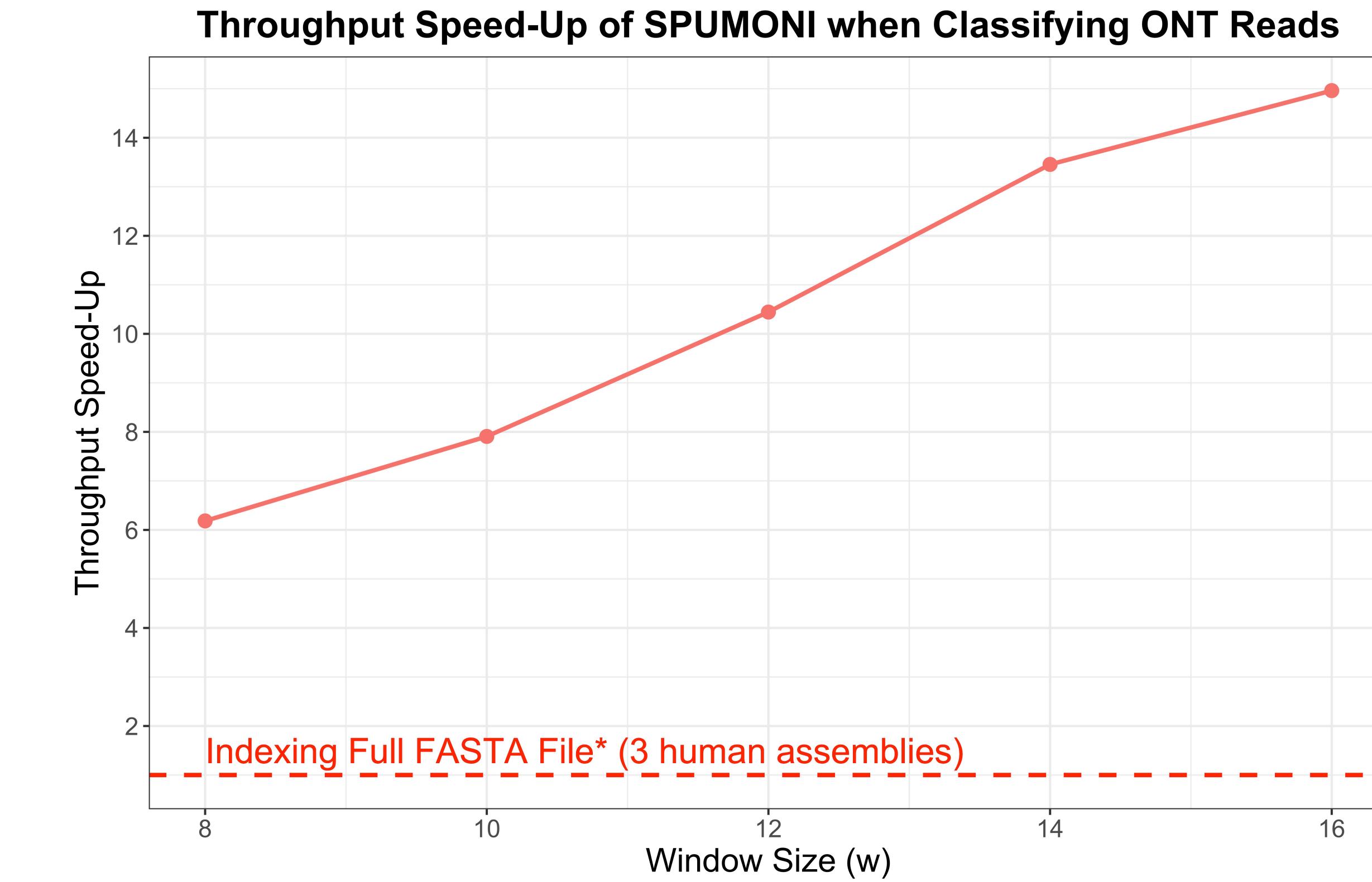
► **Question:** Can we index larger databases more efficiently than previously?

- For reference previously, SPUMONI could **index 3 human genomes in 18 GB**
- SPUMONI was **~2x faster than minimap2** at classifying ONT reads using the 3 human genome index.

► **Answer:** *Built an index over 12 human assemblies, and compared to our previous throughput ...*



Index is ~7 GB, and it has 4X as many genomes!



Results - Extending to Multiple Classes

► Question: Can we use matching statistics to distinguish multiple classes?

- In this experiment, we simulated E. coli², Salmonella², and Human Reads¹

- Built three separate indexes ...

- ▶ 3 human genomes (~9 GB)
- ▶ 1833 E. coli genomes (~9 GB)
- ▶ 988 Salmonella genomes (~4.6 GB)

} Relatively similar sized so we can just test it with simple classification rule.

- Used a simple test of largest mean to classify read however ...

► Answer: Yes, we can. Here are the confusion matrices for classifying short and long reads ...

Classifying Short Reads

True Class	Predicted Class		
	E. coli	Human	Salmonella
E. coli	24,588	39	373
Human	14	24,981	5
Salmonella	484	0	24,516

Classifying Long Reads

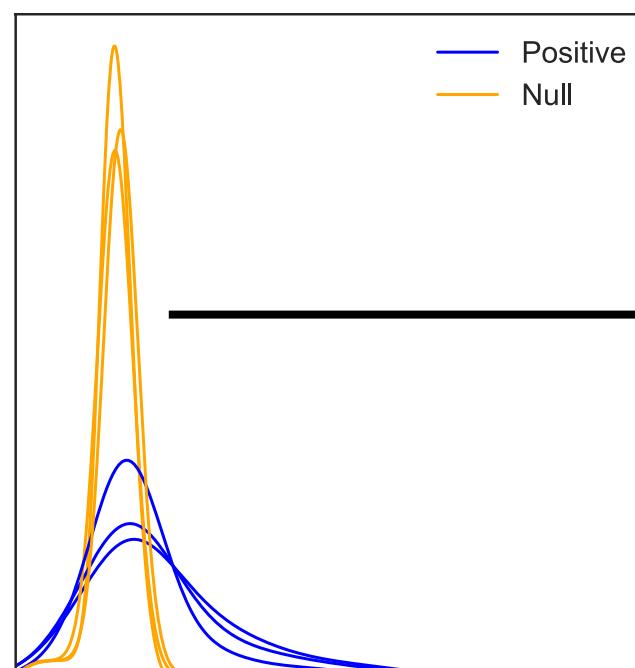
True Class	Predicted Class		
	E. coli	Human	Salmonella
E. coli	24,459	46	495
Human	30	24,946	24
Salmonella	272	17	24,711

¹ Nurk S, Koren S, Rhie A, Rautiainen M, et al. The complete sequence of a human genome. bioRxiv, 2021.

² O'Leary NA, et al. Reference sequence (RefSeq) database at NCBI: current status, taxonomic expansion, and functional annotation. Nucleic Acids Res. 2016.

Key Takeaways

- ▶ SPUMONI is a rapid tool for binary read classification that uses a read's MS or PMLs to classify it.
 - Use of the **minimizers** and a **promoted alphabet** allows SPUMONI to index large databases more efficiently.
- ▶ In a host-depletion scenario, where SPUMONI indexed human genomes ...
 - Previously, SPUMONI was ~2X faster at classifying than minimap2 (3 human genomes).
 - Using the new indexing approach and **4x as many human genomes**, SPUMONI can classify reads **6-15X faster than previously with only 3 human genomes**.
- ▶ We envision in the future we can classify metagenomic reads robustly **using distributions of matching statistics** to account for database growth and variable-sized classes.



“Null” distribution allows the **notion of significance to be a function of database sequences**, and the **error rate of the query read**.

Thank you!

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GitHub: <https://github.com/oma219/spumoni>

Acknowledgements:

- ▶ Special thanks to Massimiliano Rossi, Daniel N. Baker, & Ben Langmead for assistance on project.
- ▶ Thanks to Sam Kovaka, Michael C. Schatz, Travis Gagie, Christina Boucher for help & assistance on the project
- ▶ Thanks to Nae-Chyun Chen, Taher Mun, Kathleen Newcomer, Anna Liebhoff, Dominik Kempa, Mao-Jan Lin and Kavya Vaddadi

Funding:



**NSERC
CRSNG**