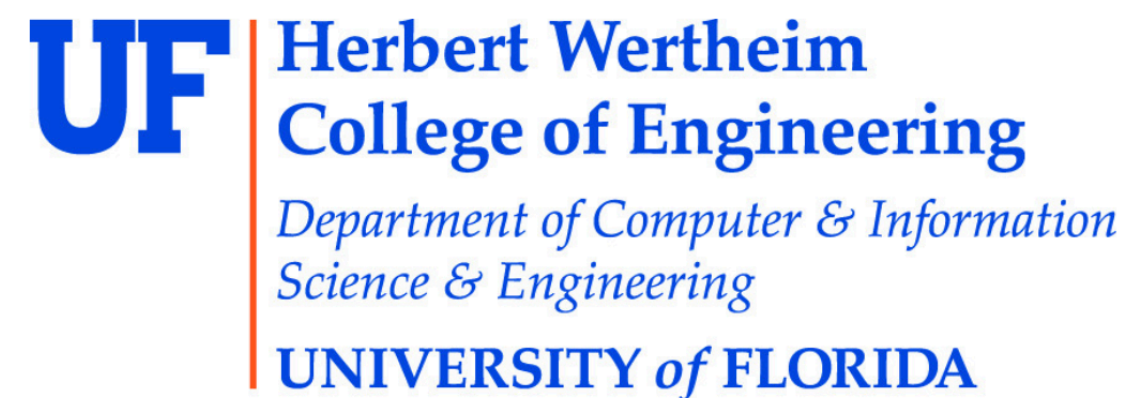


Pan-genomic indexes for robust classification of nanopore and metagenomic reads

Omar Ahmed¹, Massimiliano Rossi², Sam Kovaka¹, Michael C. Schatz¹, Travis Gagie³,
Christina Boucher², and Ben Langmead¹

Genome Informatics
November 3-5, 2021



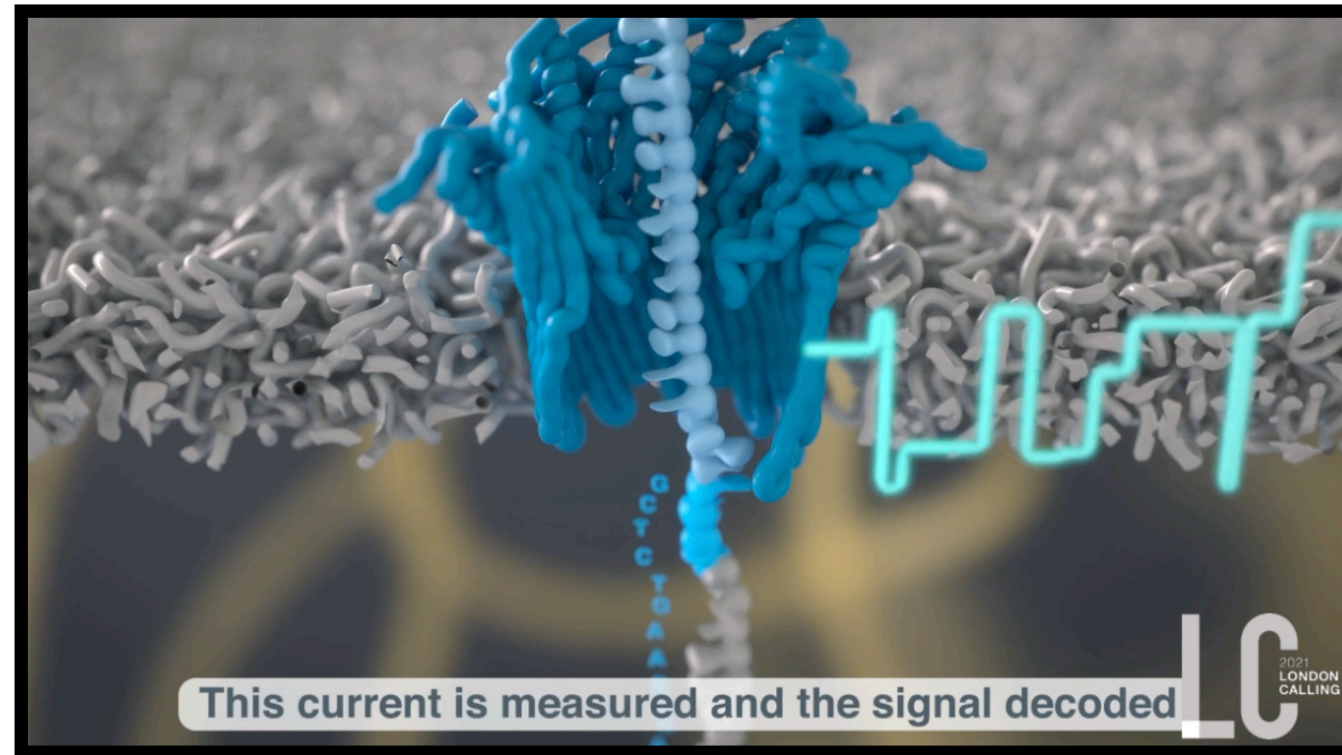
¹ Department of Computer Science, Johns Hopkins University, Baltimore, MD, USA

² Department of Computer and Information Science and Engineering, University of Florida, Gainesville, FL, USA

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

Overview of Presentation

- ▶ Development of SPUMONI for classification of nanopore reads

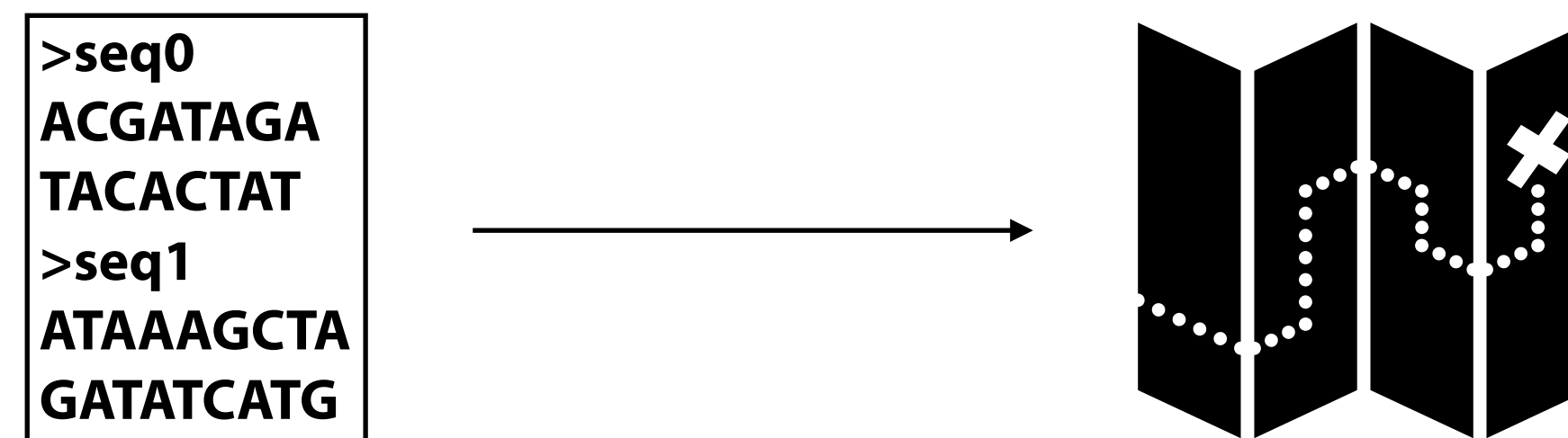


$T = a \ c \ g \ g \ c \ t \ a \ c \ a \ t \ a \ \$$

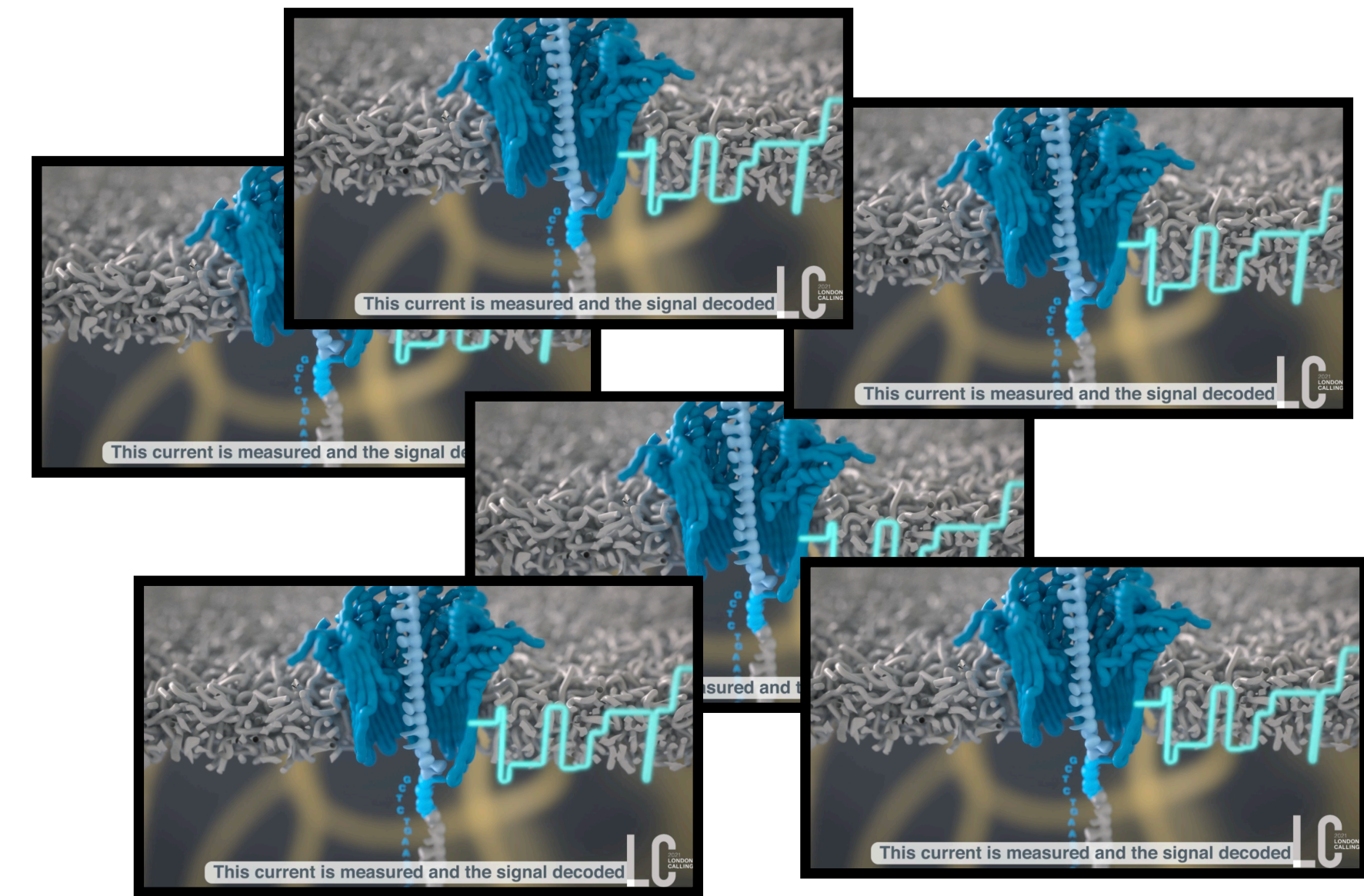
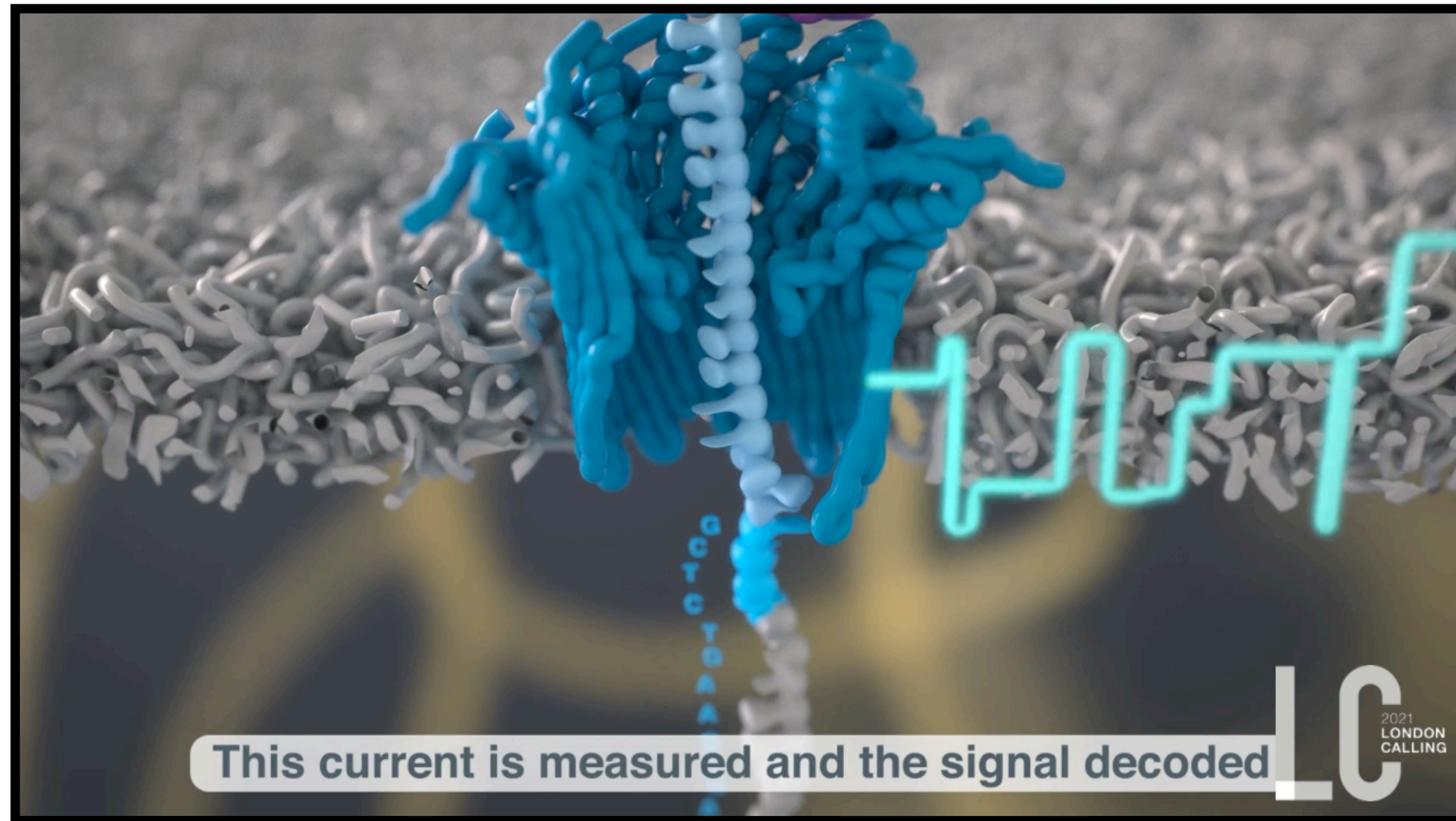
$P = t \ a \ c \ g \ g \ t \ a$

$M = 3 \ 4 \ 3 \ 2 \ 1 \ 2 \ 1$

- ▶ 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

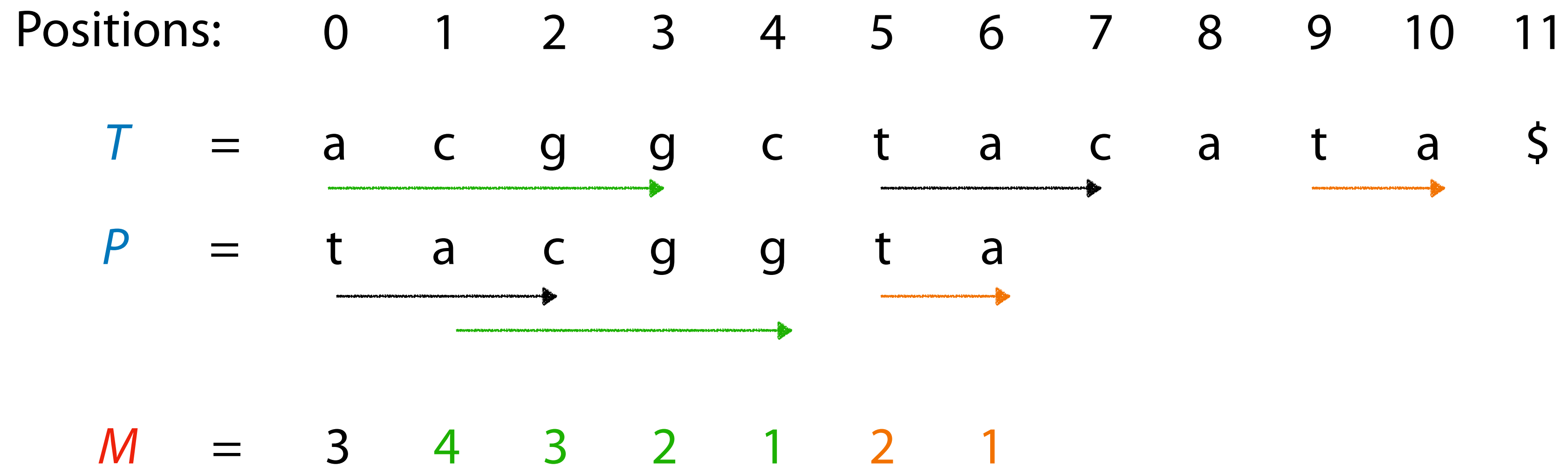
Matching Statistics

▶ 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

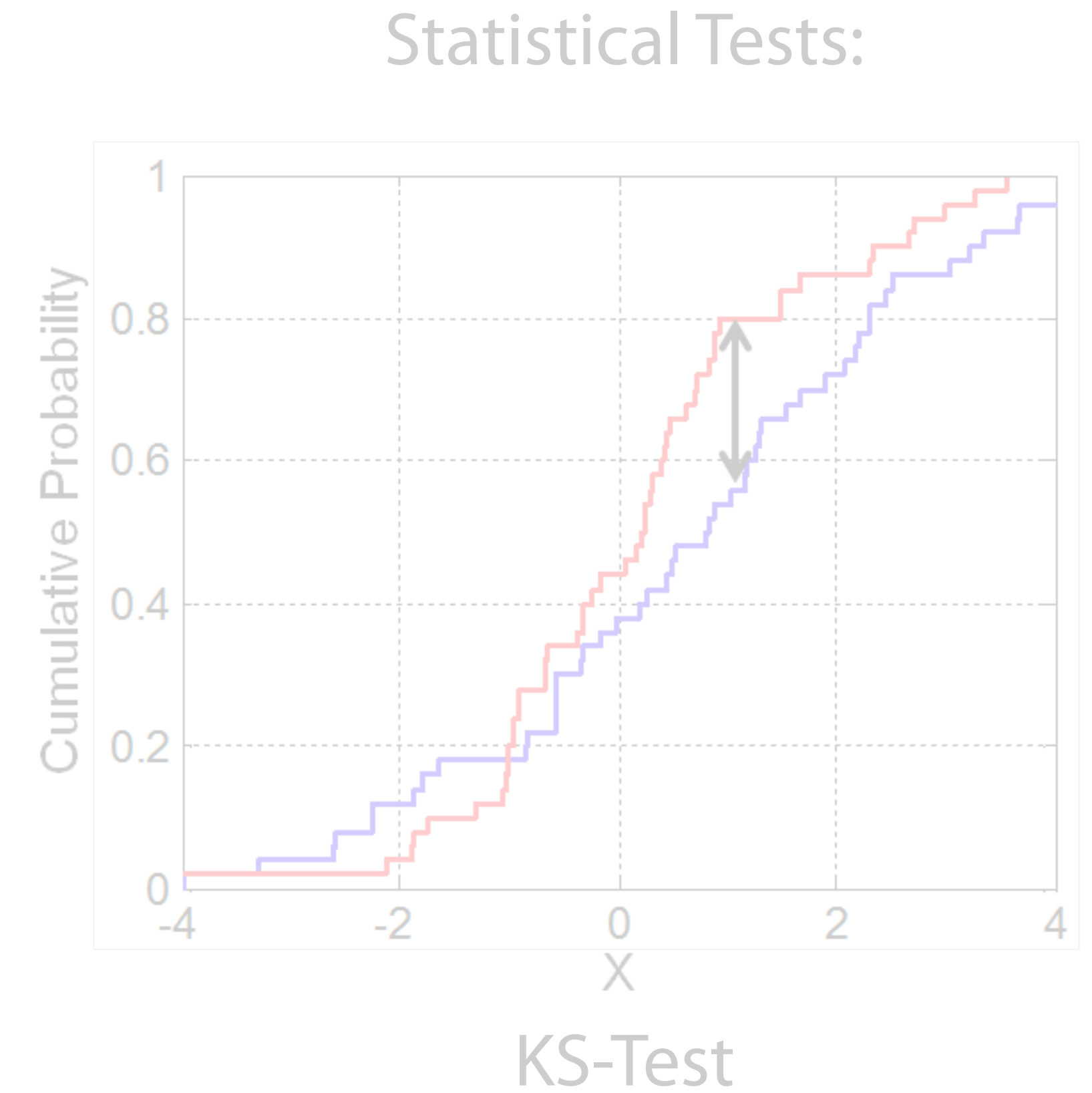
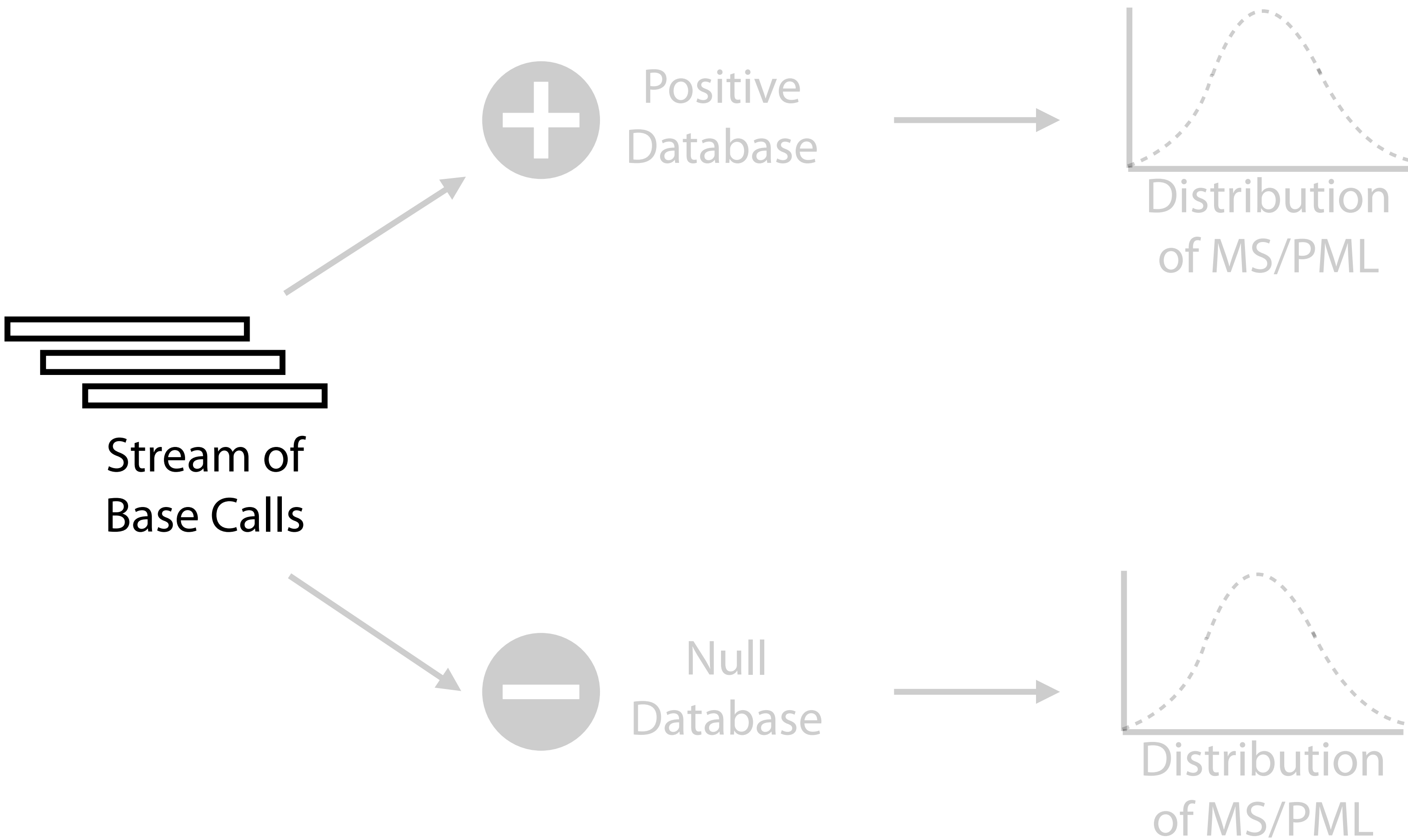
○ Let T be a text of length n , and P be a pattern of length m

▶ Think of matching statistics like they are half MEMs (Maximally Exact Matches)

○ Example:



SPUMONI Approach



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!

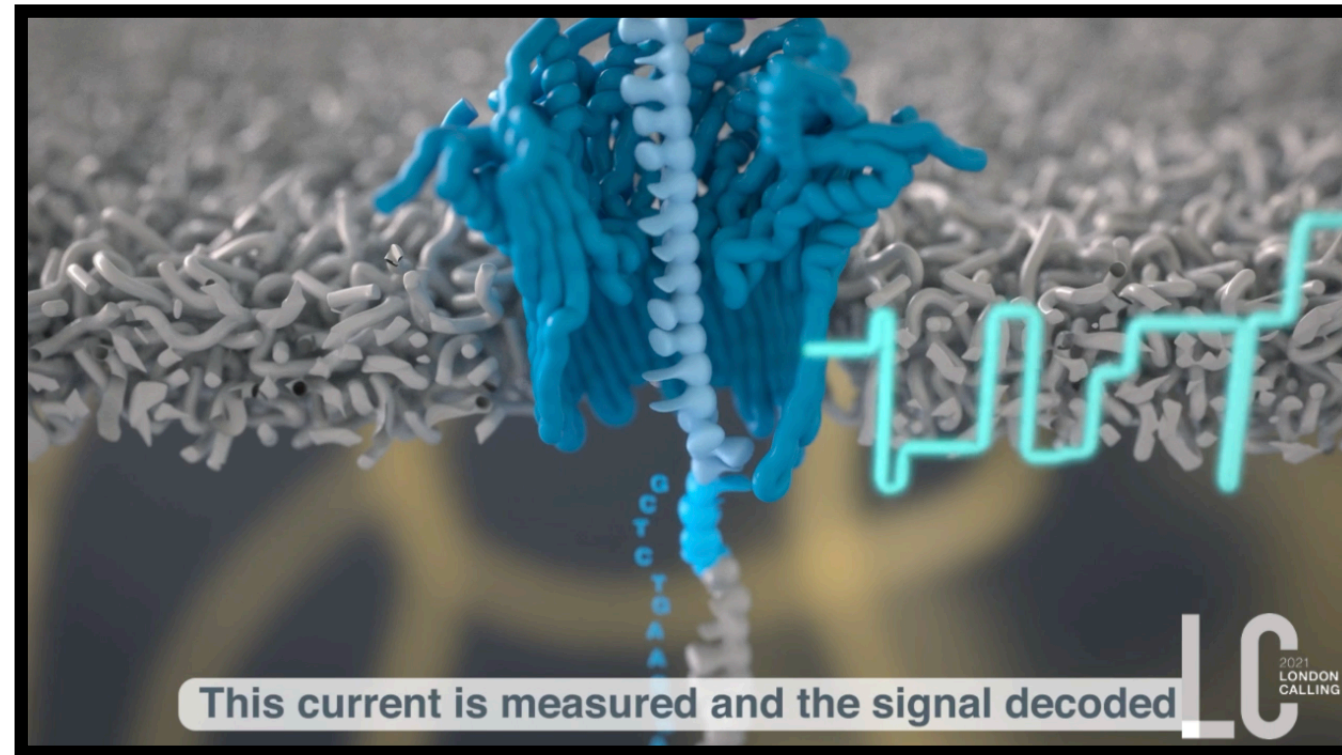
► For details on additional specifics of method and results:

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

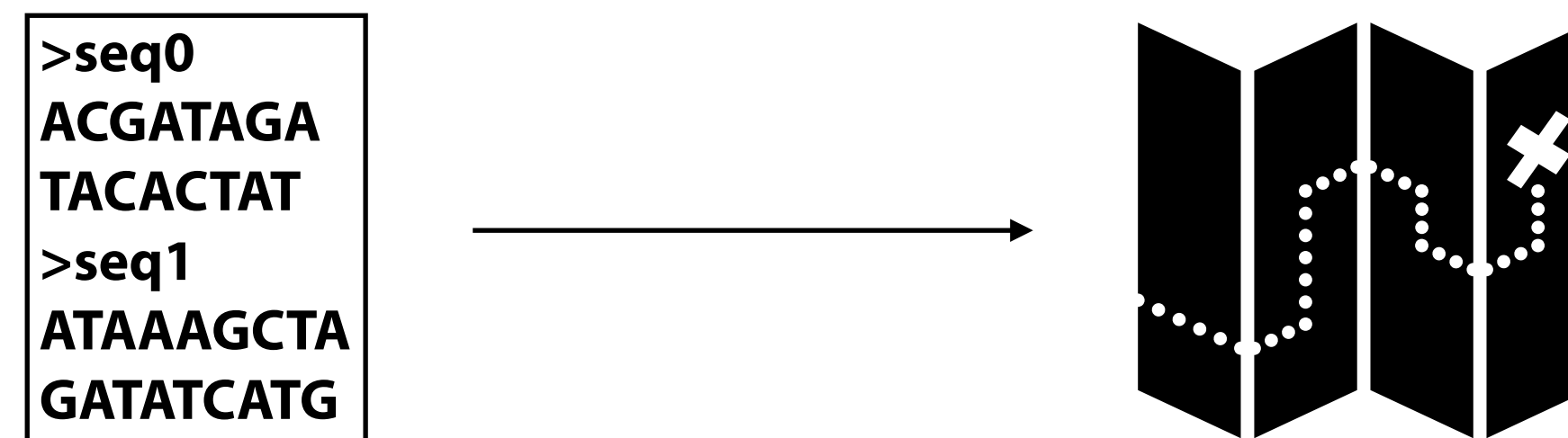
Overview of Presentation

- ▶ Development of SPUMONI for classification of nanopore reads



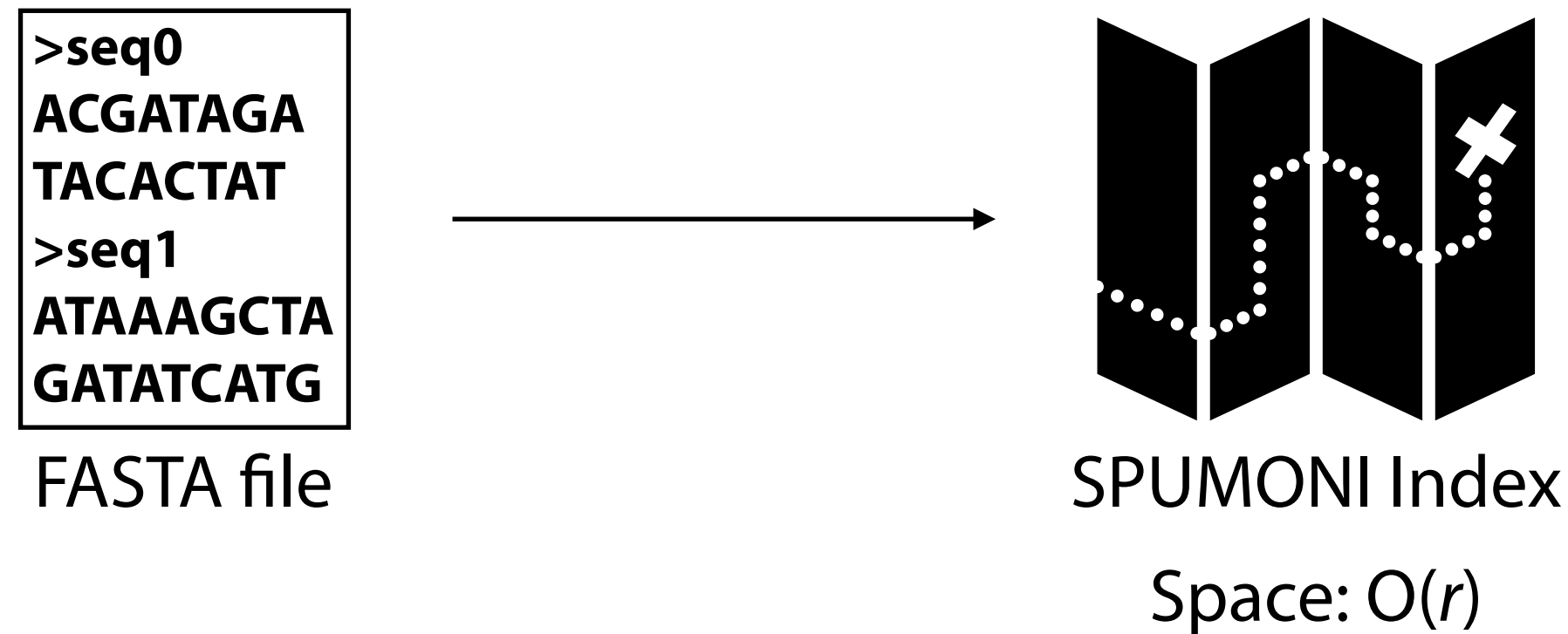
T = a c g g c t a c a t a \$
P = t a c g g t a
M = 3 4 3 2 1 2 1

- ▶ Scaling SPUMONI to handle larger pan-genomes more efficiently



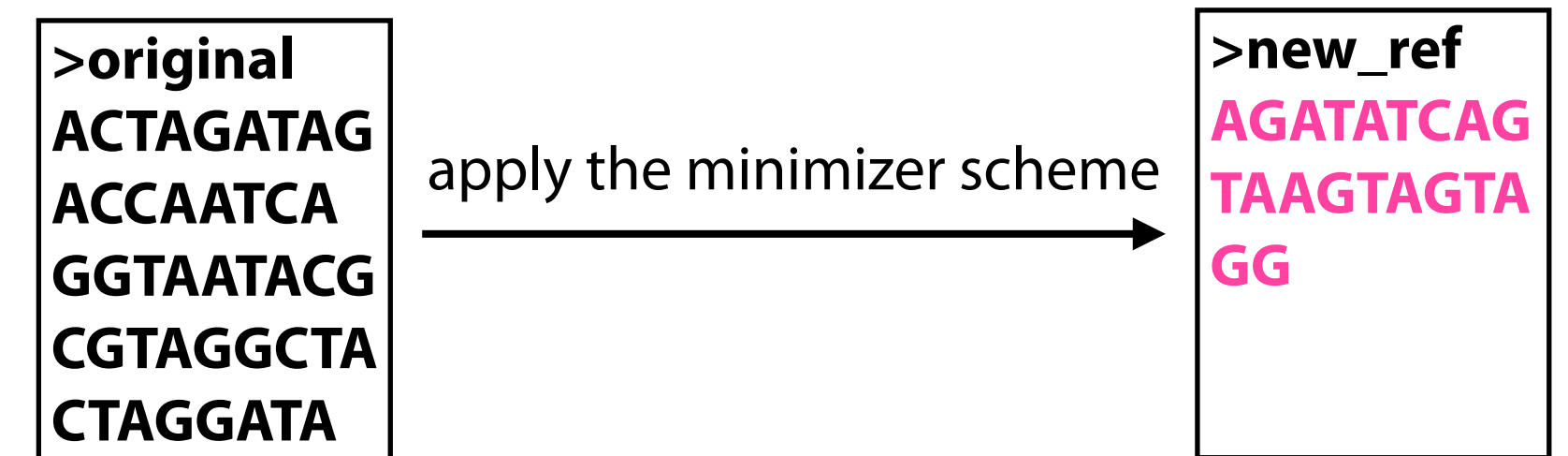
Speeding Up SPUMONI

► 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.

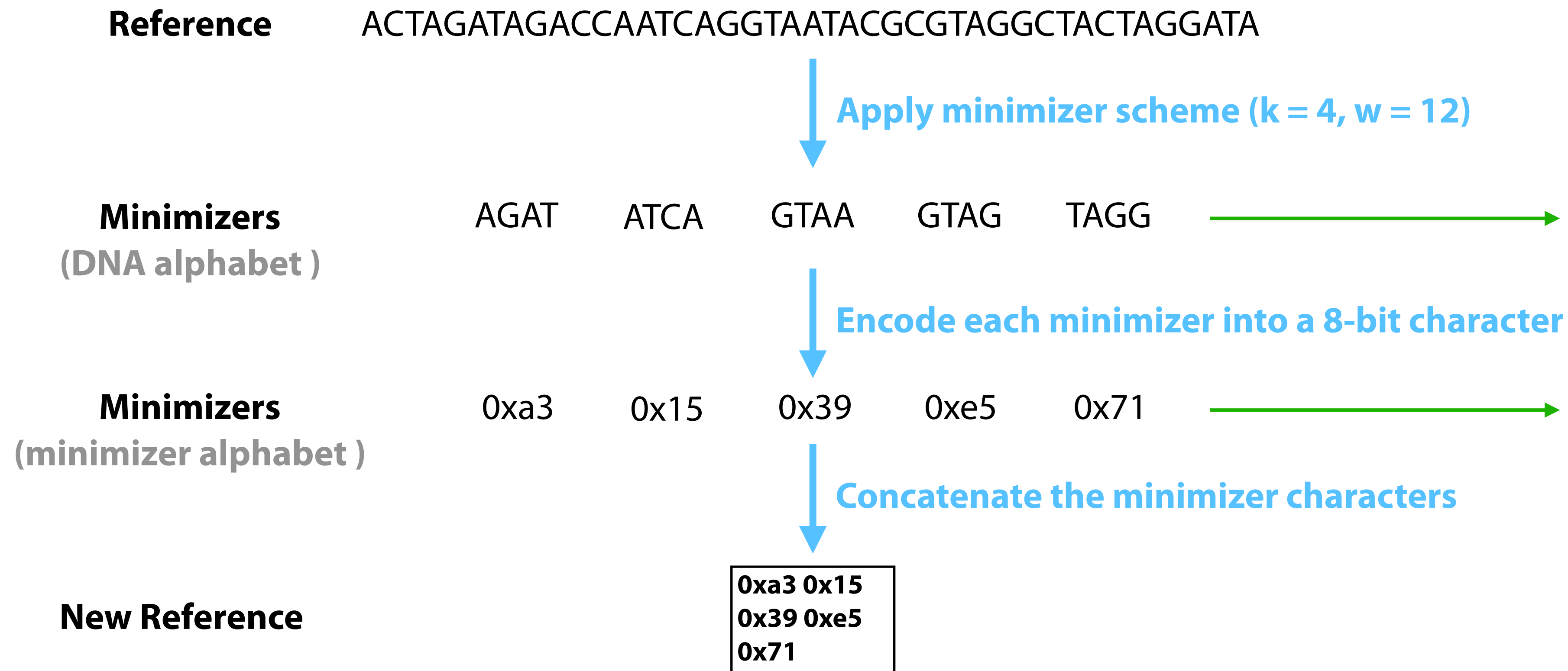
¹ Roberts, M., Hayes, W., Hunt, B. R., Mount, S. M., & Yorke, J. A. (2004). Reducing storage requirements for biological sequence comparison. *Bioinformatics*, 20(18), 3363-3369.

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 ...



Impact of the Alphabet Promotion

```
seq.length = 5 * 4  
for i = 1 to i = seq.length {  
    // Compute MS at position i  
}
```

```
seq.length = 5  
for i = 1 to i = seq.length {  
    // Each iteration covers 4  
    // characters  
}
```

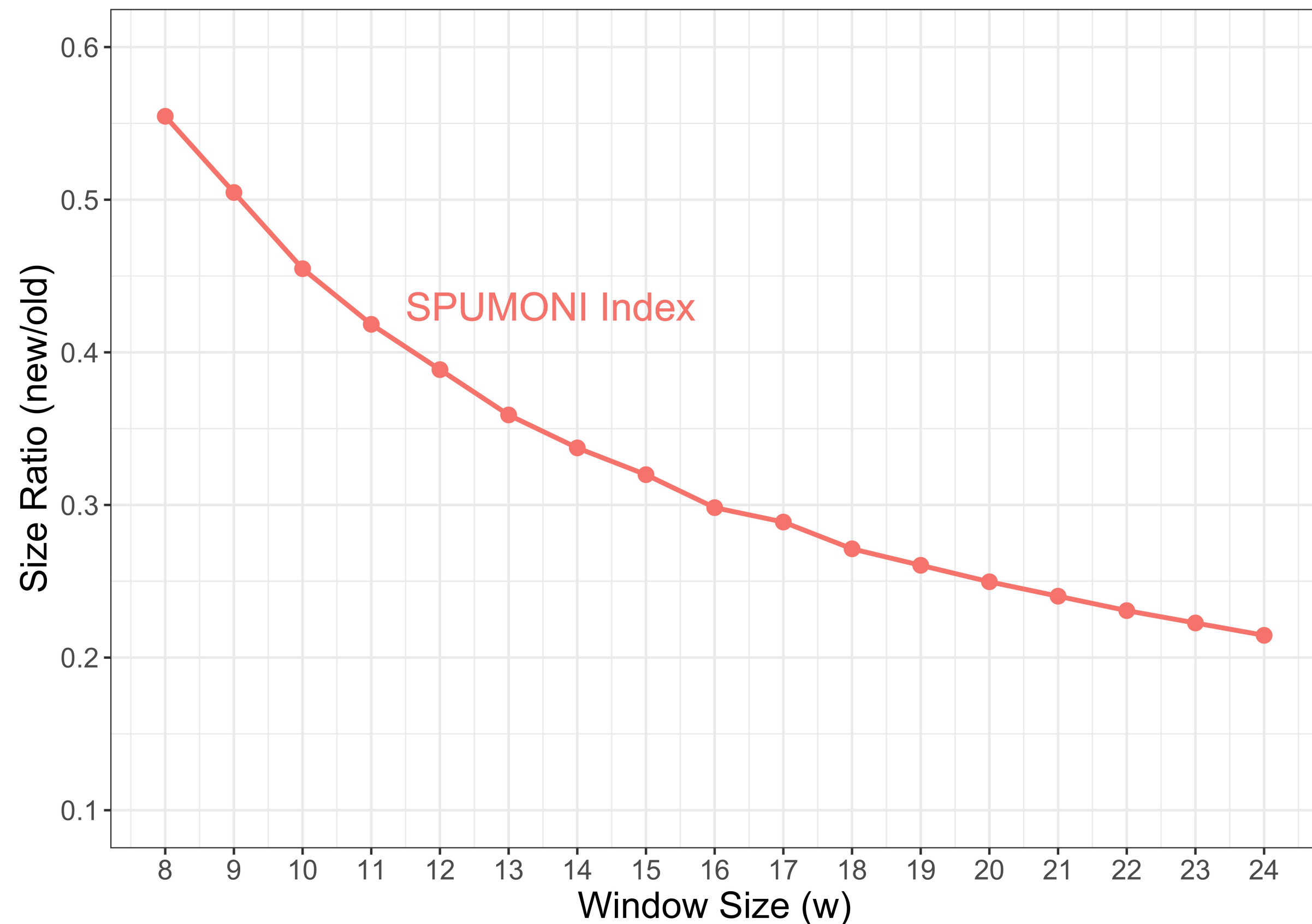
Combining the minimizer scheme with alphabet promotion leads to even smaller references.

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*.

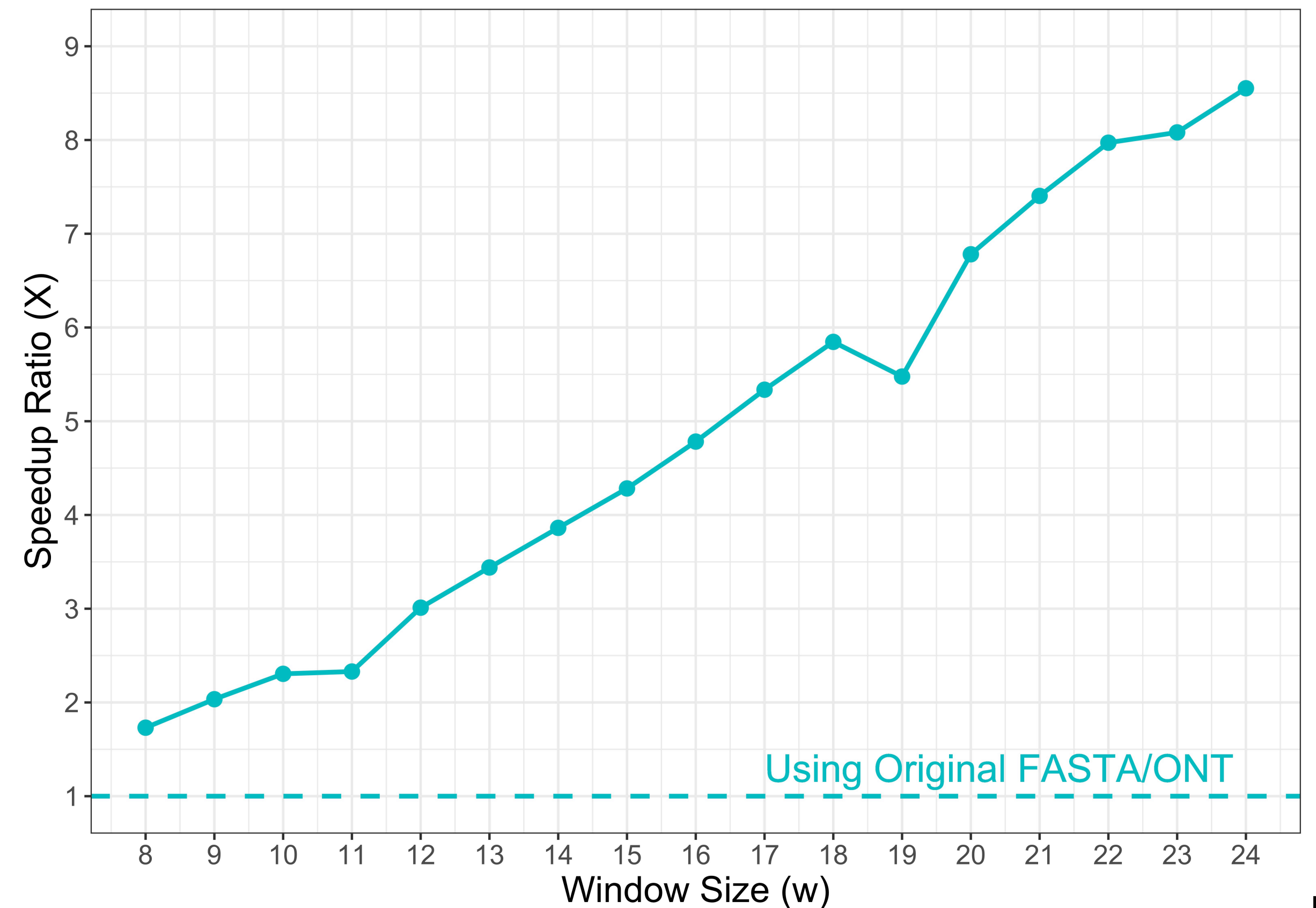
Results - Using Minimizer-Based References

- ▶ **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 ...*

File Size Comparisons After Applying Minimizer Digest



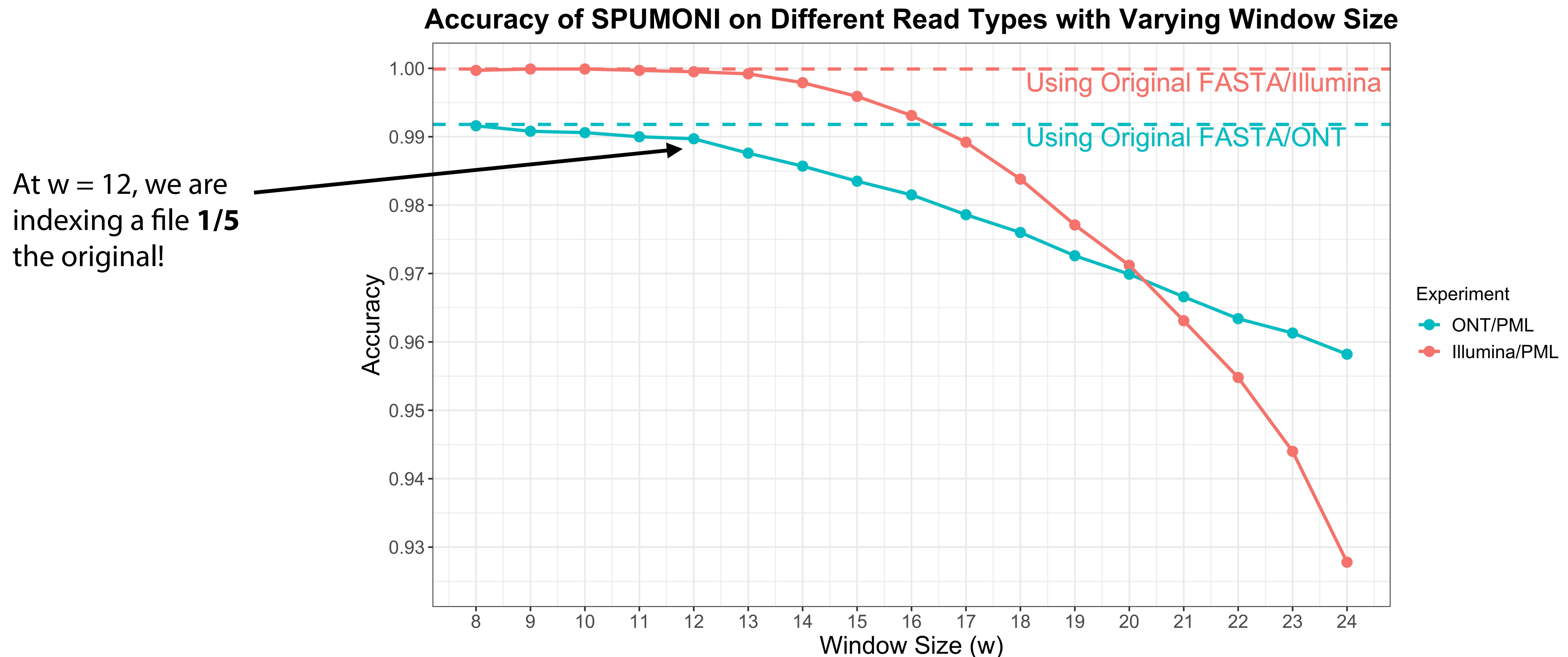
Speedup Ratio for SPUMONI Using Minimizer-Based References



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 ...*



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.

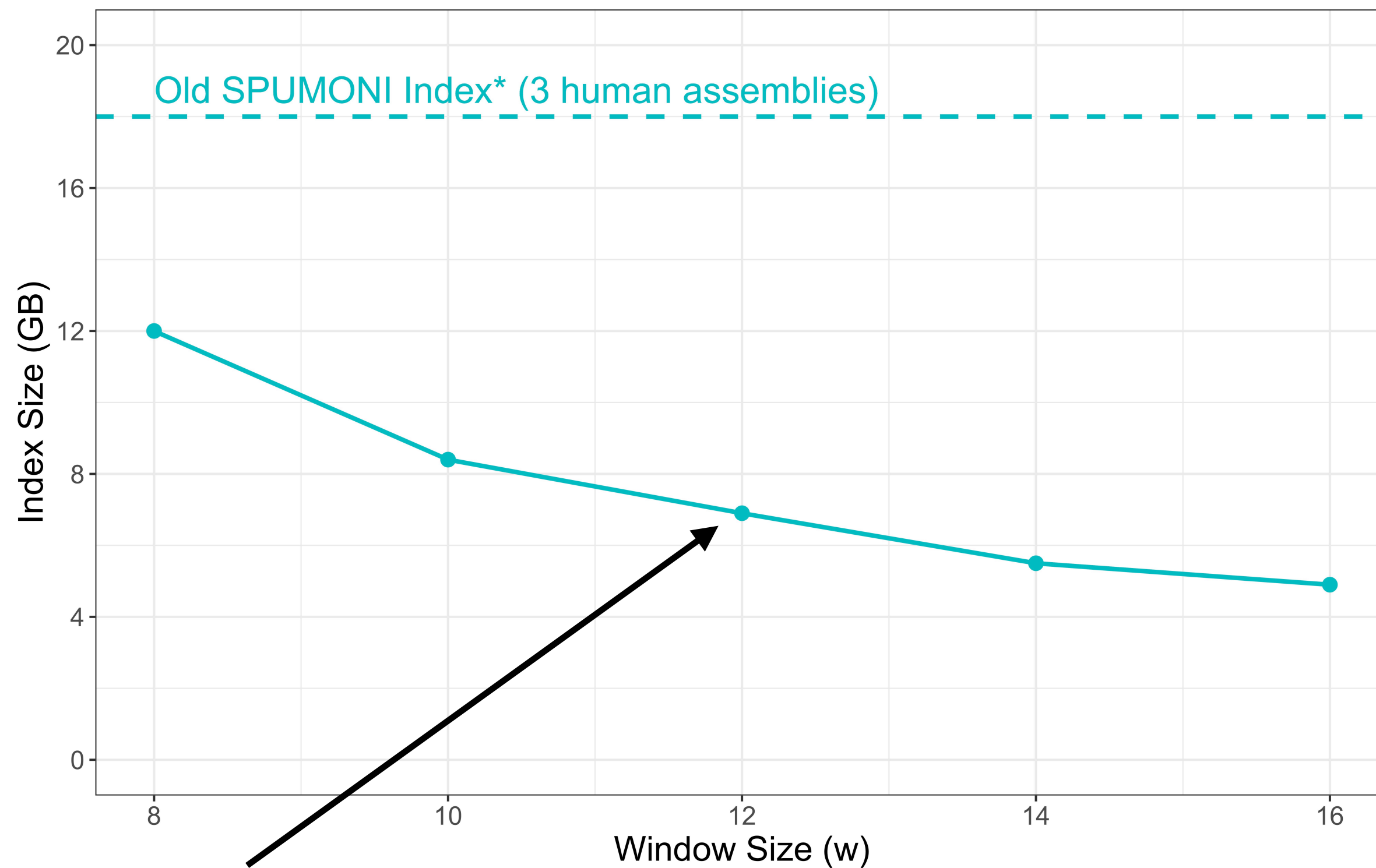
Results - Indexing 12 Human Assemblies

► **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.

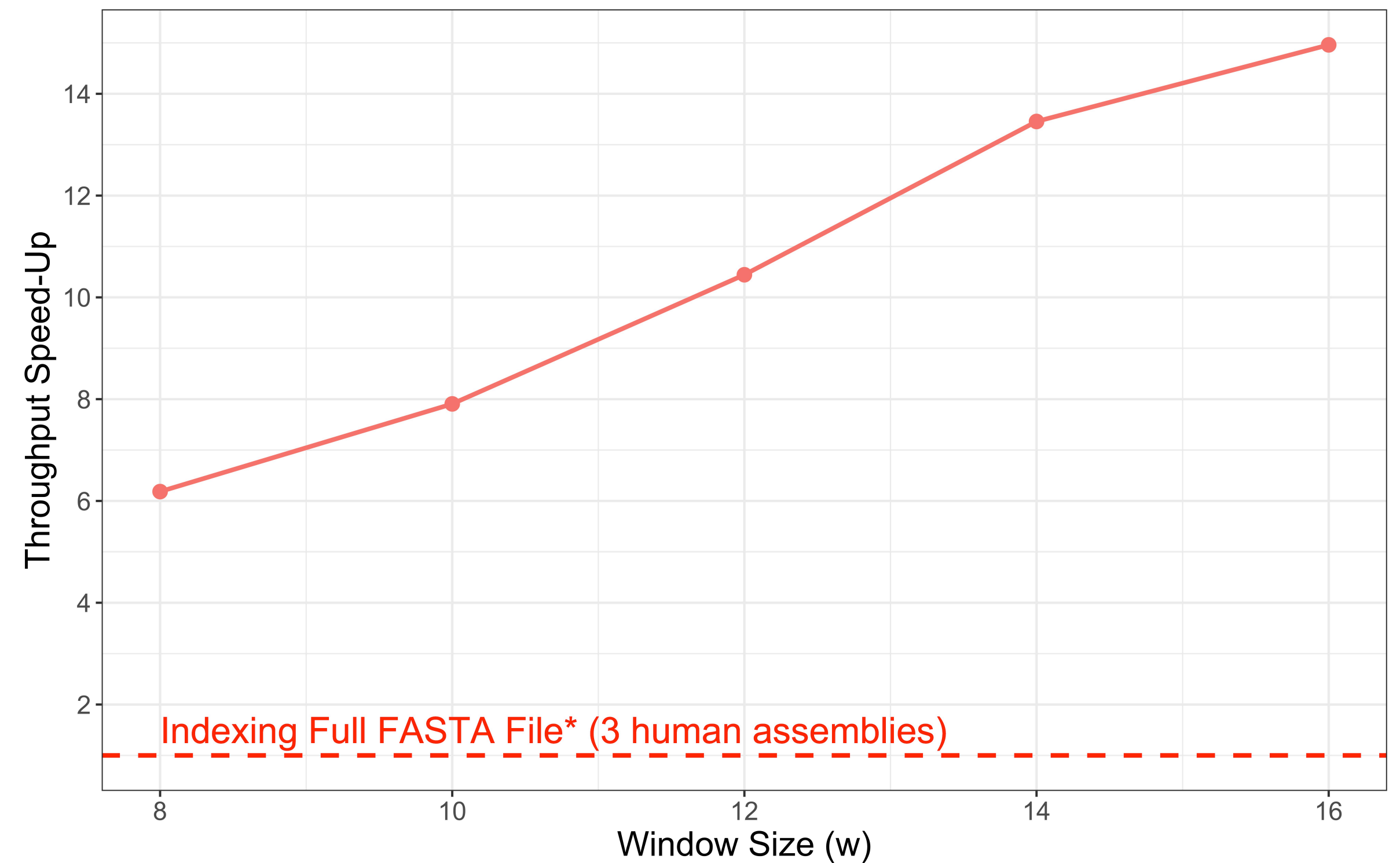
► **Answer:** *Built an index over 12 human assemblies, and compared to our previous throughput ...*

SPUMONI Index Sizes for 12 Human Assemblies



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

Throughput Speed-Up of SPUMONI when Classifying ONT Reads



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

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

Classifying Long Reads

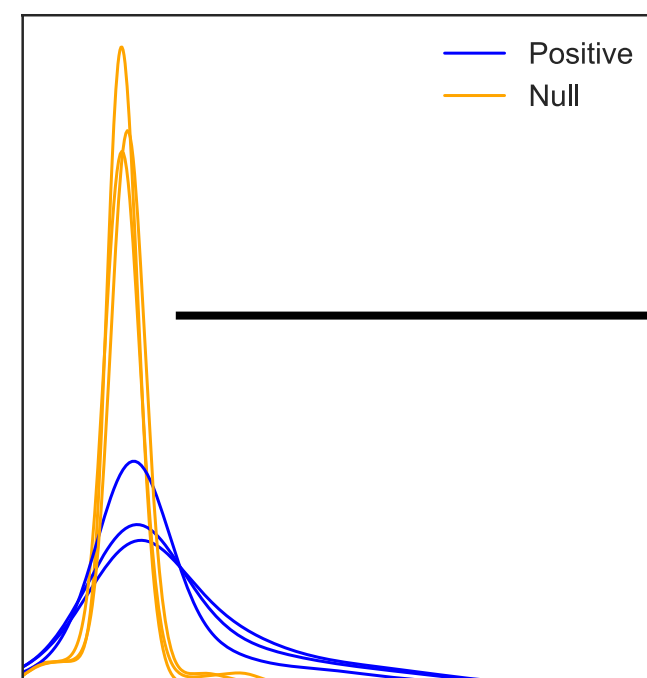
		Predicted Class		
		E. coli	Human	Salmonella
True Class	E. coli	24,459	46	495
	Human	30	24,946	24
	Salmonella	272	17	24,711

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!

Contact: oahmed6@jhu.edu

Twitter: [@oyfahmed](https://twitter.com/oyfahmed)

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