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# **Top- $k$ Frequent Patterns in Streams and Parameterized-Space LZ Compression**

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dortmund



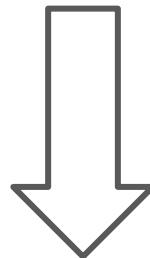
Università  
Ca'Foscari  
Venezia

# **Lempel-Ziv (LZ) Compression**

panaman\_ananas\_banana\_pancake

# Lempel-Ziv (LZ) Compression

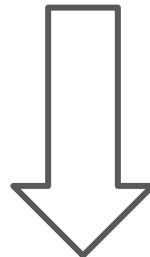
panam**an**\_ananas\_banana\_pancake



panam**(4, 2)**\_ananas\_banana\_pancake

# Lempel-Ziv (LZ) Compression

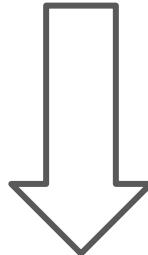
panaman\_ananas\_banana\_pancake



panam(4, 2)\_ananas\_banana\_pancake

# Lempel-Ziv (LZ) Compression

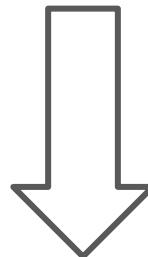
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panam(4,2)-(7,3)nas\_banana\_pancake

# Lempel-Ziv (LZ) Compression

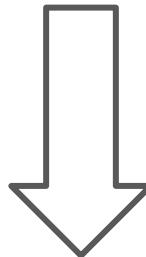
panaman\_ananas\_banana\_pancake



panam(4,2)-(7,3)(2,2)s\_b(8,5)-(22,3)cake

# Lempel-Ziv (LZ) Compression

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panam(4,2)\_-(7,3)(2,2)s\_b(8,5)\_-(22,3)cake

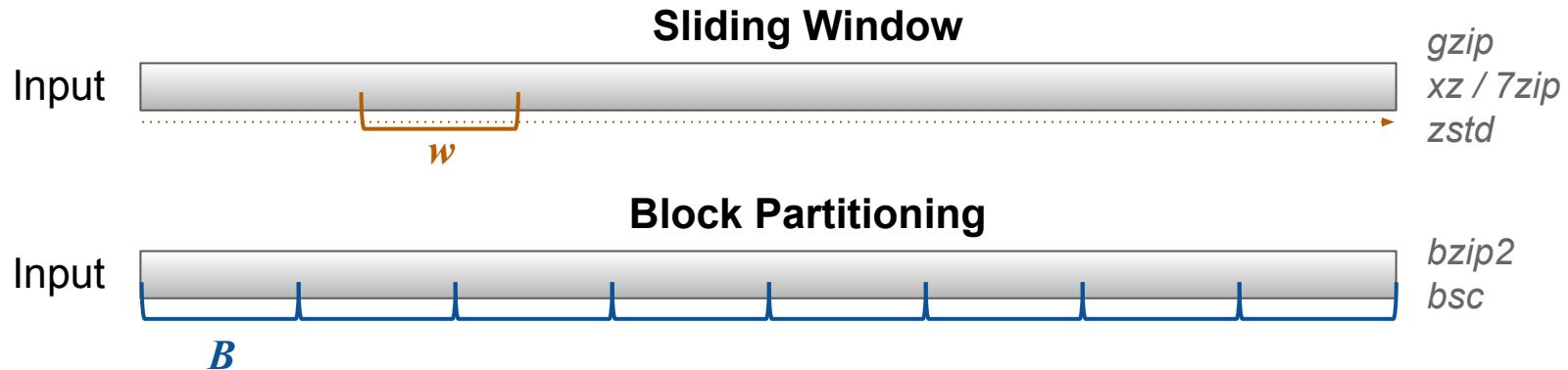
**LZ references**

# **Compression in parameterized space**

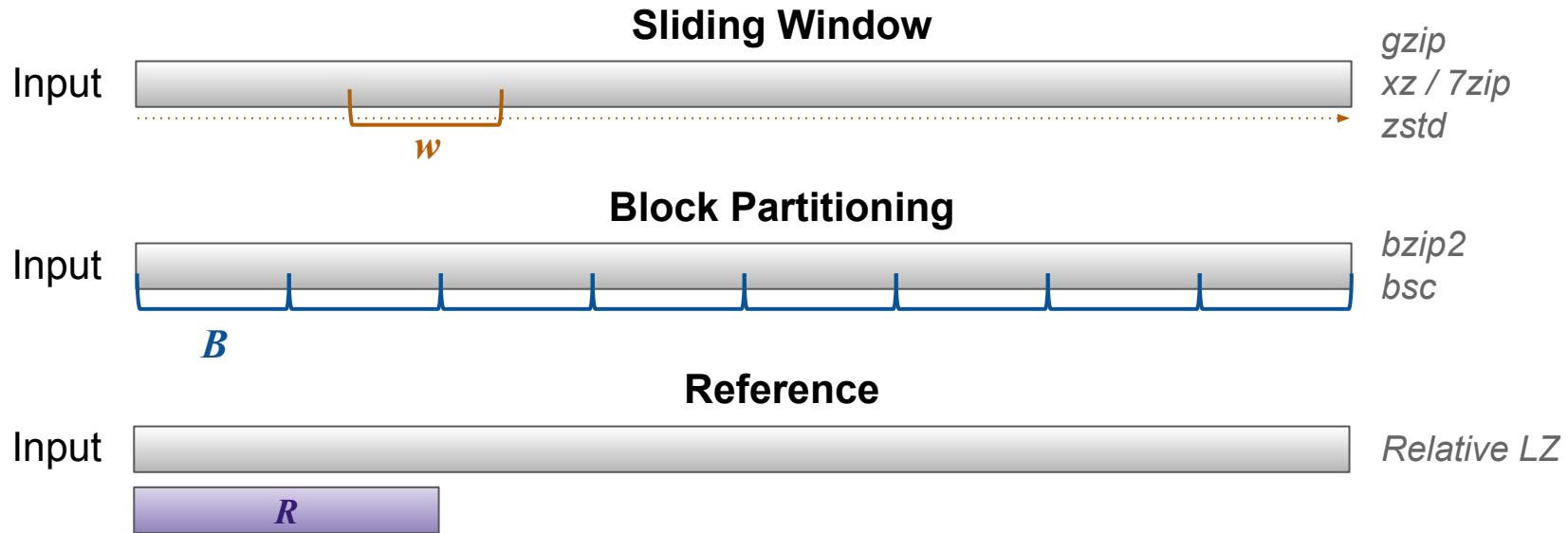
# Compression in parameterized space



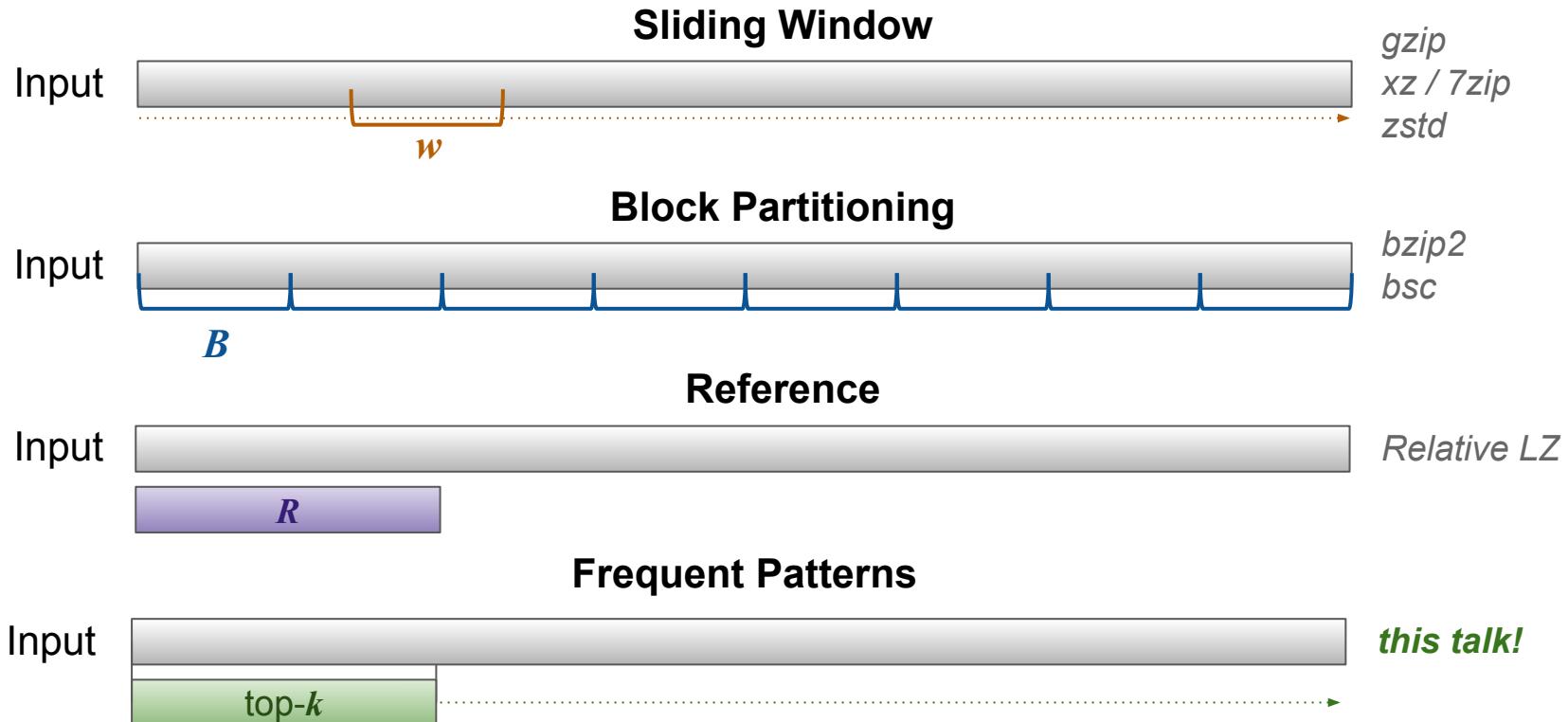
# Compression in parameterized space



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# Frequent Patterns

→ We look for frequent **patterns** in a stream  $S$  of characters

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# Compressing with Frequent Patterns

**Idea:**

- Estimate online which  $k$  patterns are the most frequent
- Replace occurrences of frequent patterns by references

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- Replace occurrences of frequent patterns by references

→ The number of candidate patterns is quadratic in  $|S|$

# **Frequent Patterns**

versus

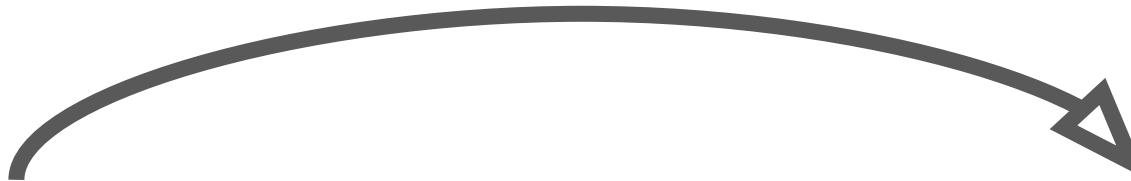
# **LZ References**

→ Frequent patterns are captured by LZ references

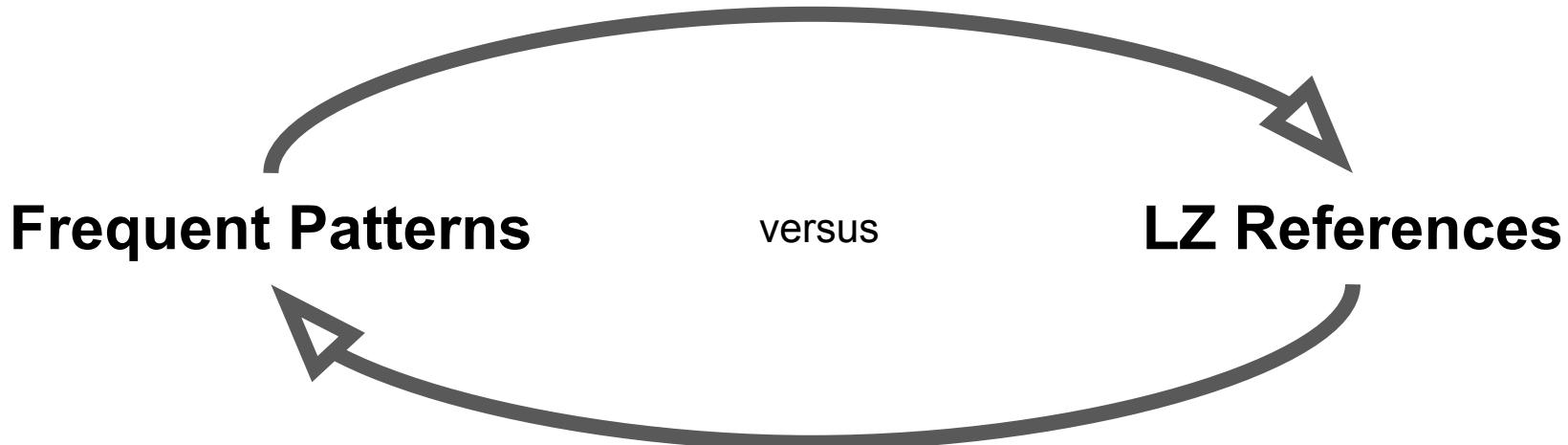
**Frequent Patterns**

versus

**LZ References**

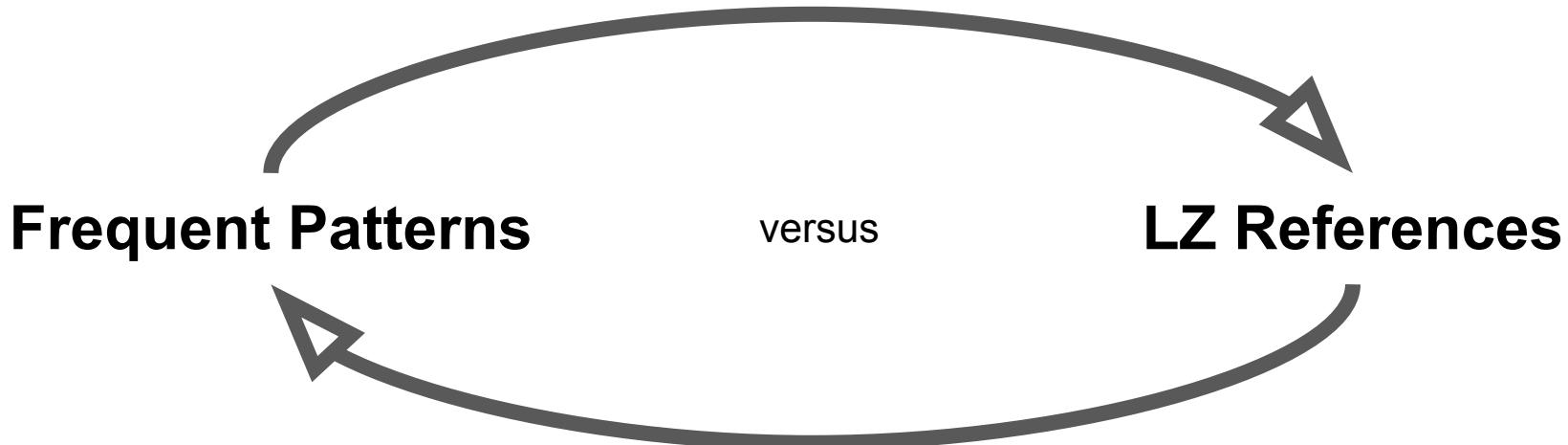


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→ The number of LZ references is bounded by  $|S|$

# top-k LZ78

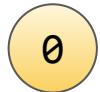
$k=6$



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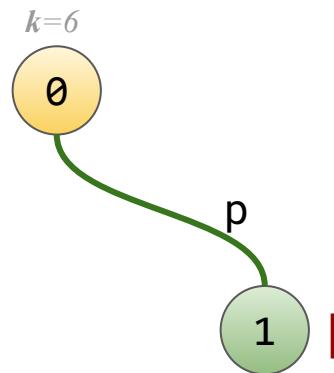
# top-k LZ78

$k=6$



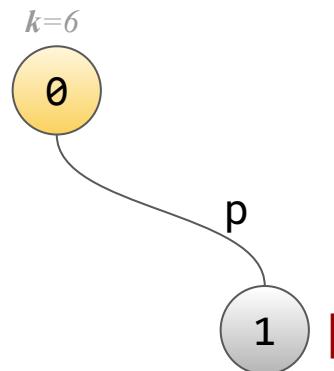
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# top-k LZ78



$(\theta, p)$

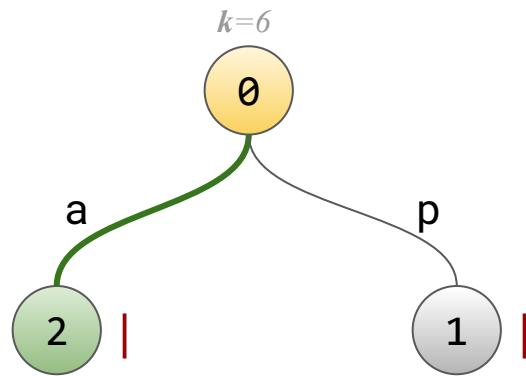
# top-k LZ78



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$(\theta, p)$

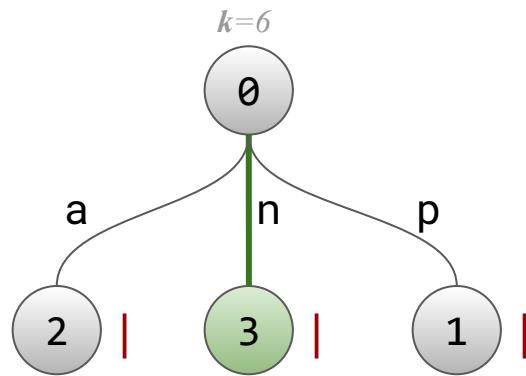
# top-k LZ78



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( $\theta$ , p)( $\theta$ , a)

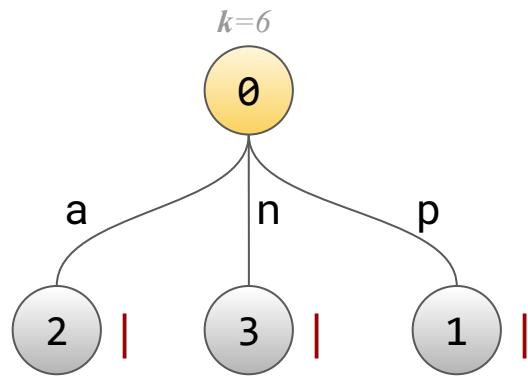
# top-k LZ78



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( $\theta, p$ ) ( $\theta, a$ ) ( $\theta, n$ )

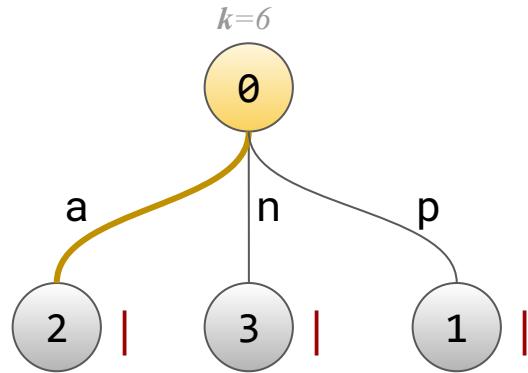
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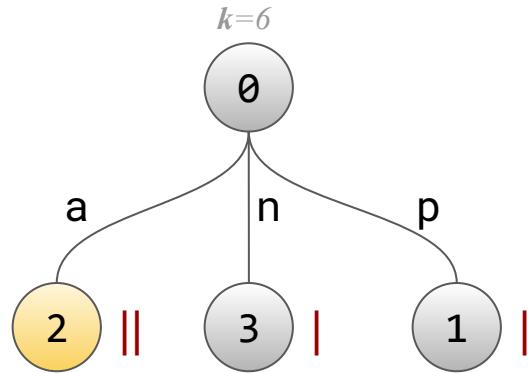
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( $\theta, p$ ) ( $\theta, a$ ) ( $\theta, n$ )

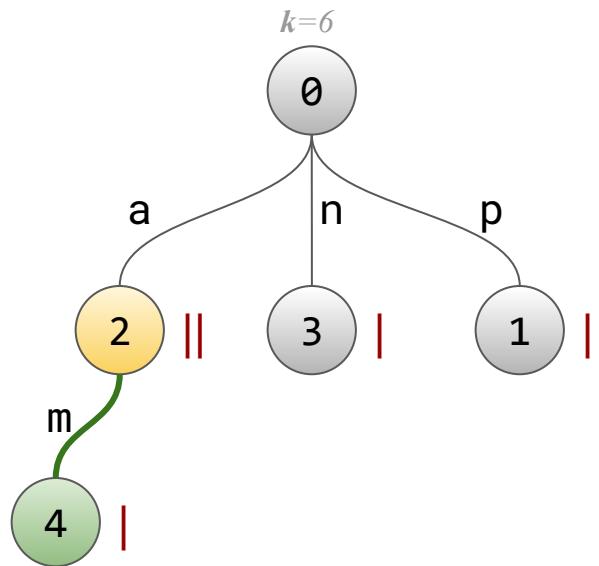
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( $\theta, p$ ) ( $\theta, a$ ) ( $\theta, n$ )

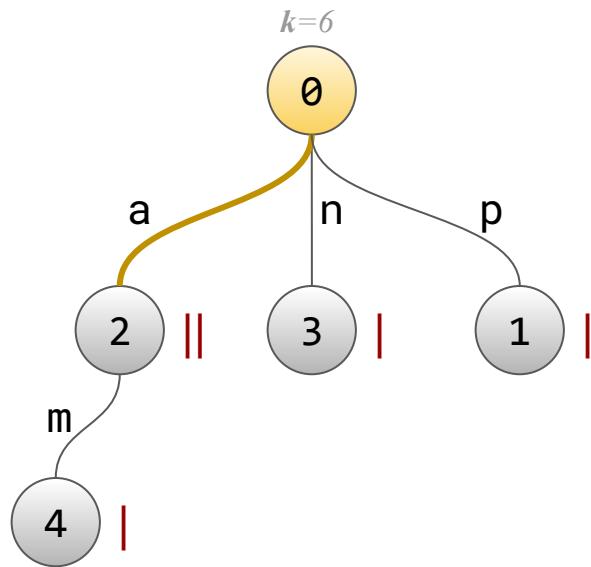
# top-k LZ78



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$(\theta, p)(\theta, a)(\theta, n)(2, m)$

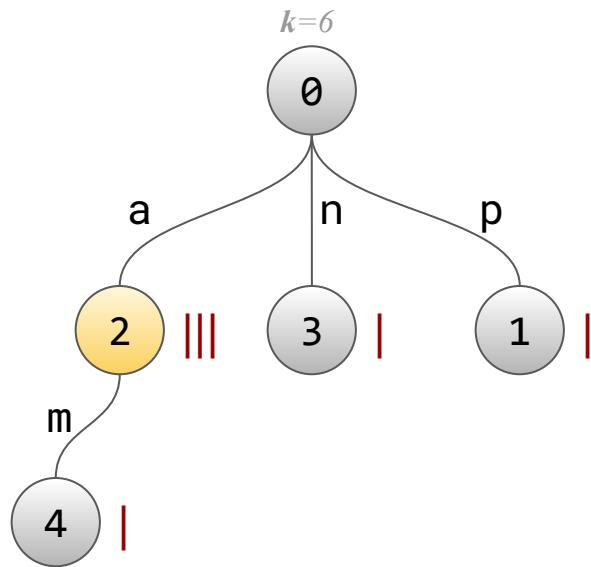
# top-k LZ78



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( $\theta, p$ ) ( $\theta, a$ ) ( $\theta, n$ ) ( $2, m$ )

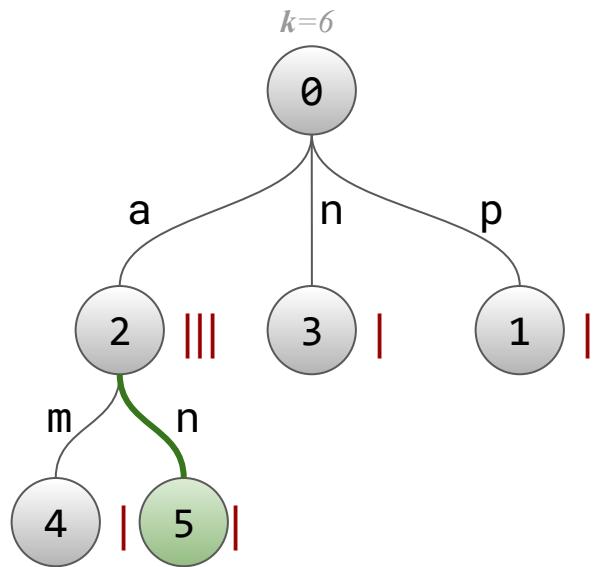
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( $\theta, p$ ) ( $\theta, a$ ) ( $\theta, n$ ) ( $2, m$ )

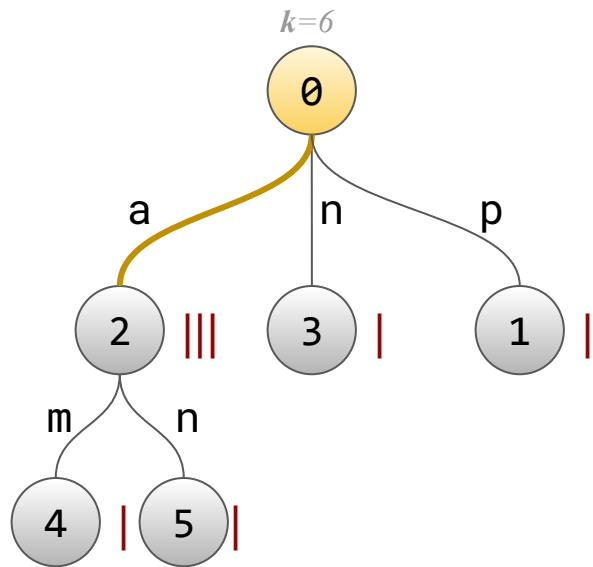
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( $\theta, p$ ) ( $\theta, a$ ) ( $\theta, n$ ) ( $2, m$ ) ( $2, n$ )

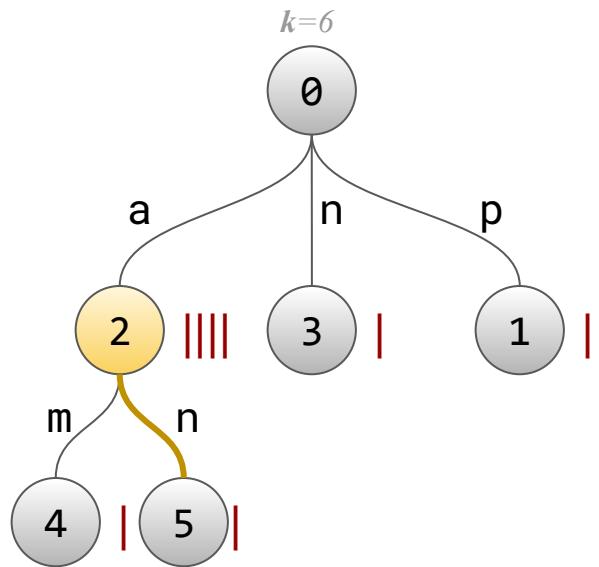
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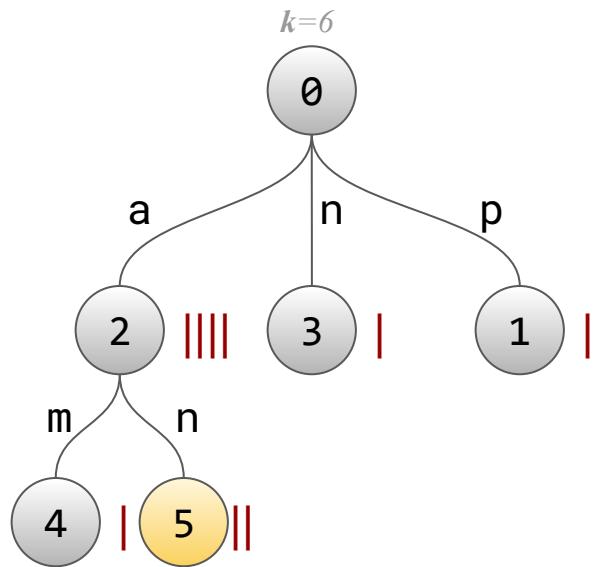
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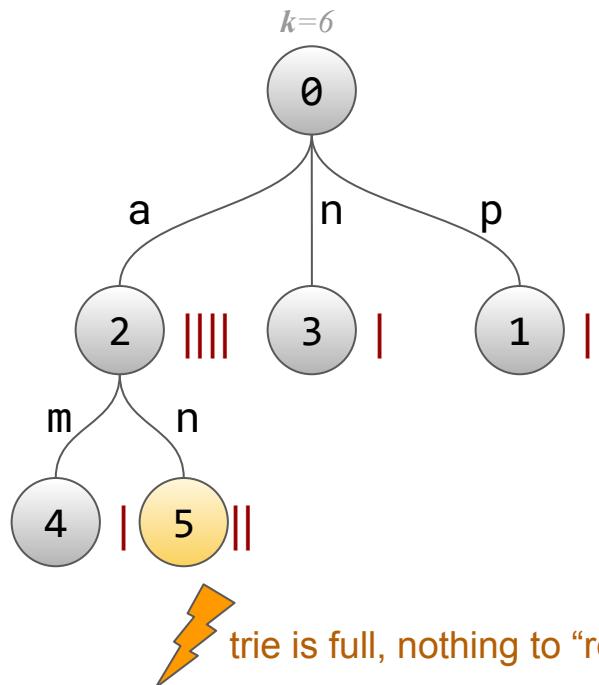
# top-k LZ78



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(0, p)(0, a)(0, n)(2, m)(2, n)

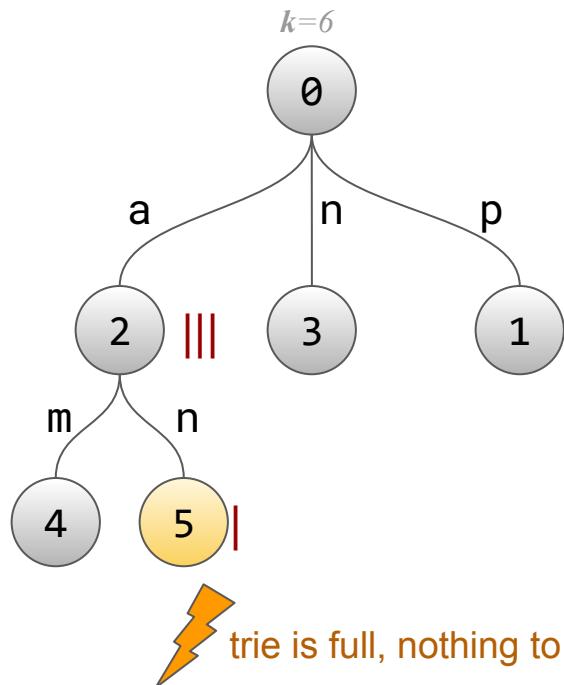
# top-k LZ78



panamanananasbananapancake

$(\theta, p)(\theta, a)(\theta, n)(2, m)(2, n)(5, a)$

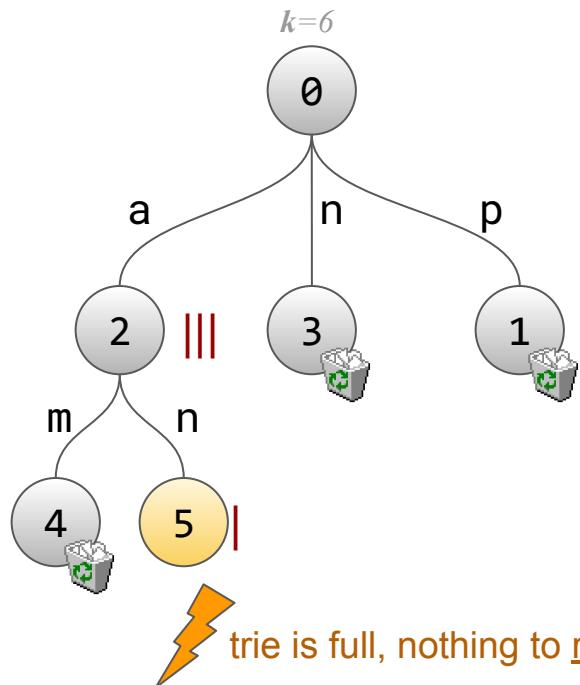
# top-k LZ78



panamanananasbananapancake

(0, p)(0, a)(0, n)(2, m)(2, n)(5, a)

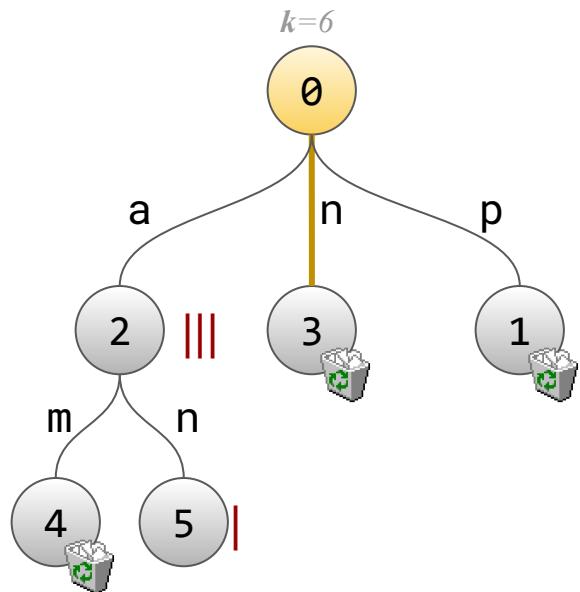
# top-k LZ78



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( $\theta, p$ ) ( $\theta, a$ ) ( $\theta, n$ ) ( $2, m$ ) ( $2, n$ ) ( $5, a$ )

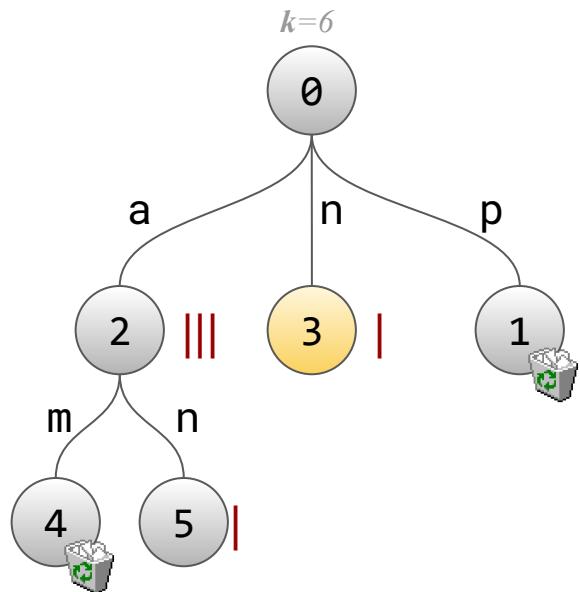
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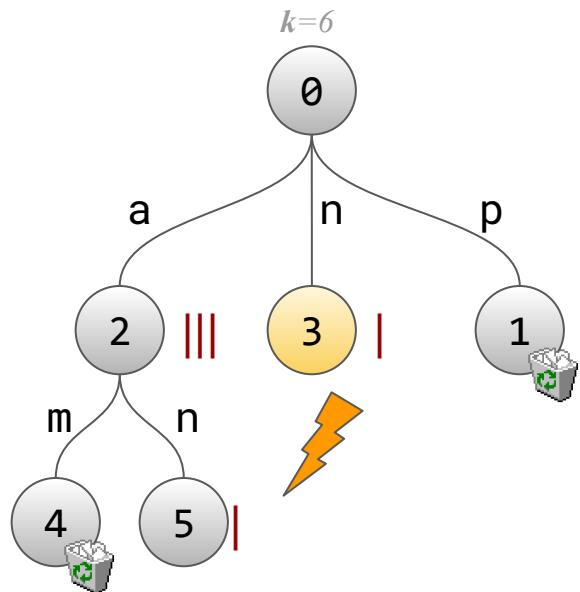
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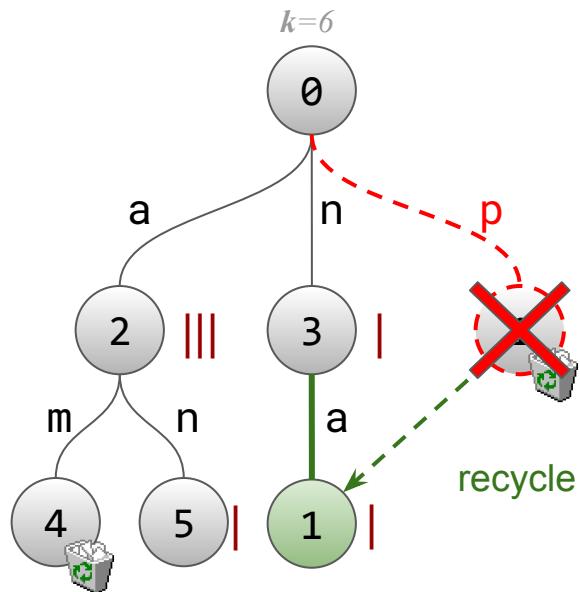
(θ, p)(θ, a)(θ, n)(2, m)(2, n)(5, a)

# top-k LZ78



(0, p)(0, a)(0, n)(2, m)(2, n)(5, a)(3, a)

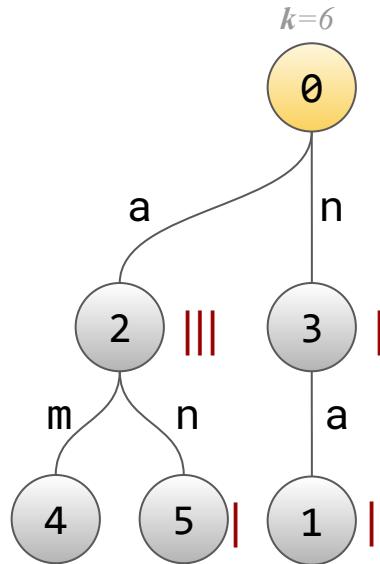
# top-k LZ78



panamanananaasbananapancake

(0, p)(0, a)(0, n)(2, m)(2, n)(5, a)(3, a)

# top-k LZ78

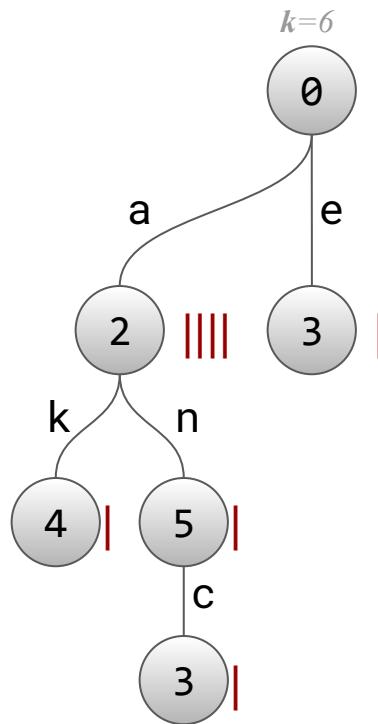


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... and so forth ...

( $\theta, p$ ) ( $\theta, a$ ) ( $\theta, n$ ) ( $2, m$ ) ( $2, n$ ) ( $5, a$ ) ( $3, a$ )

# top-k LZ78



panamanananasbananapancake

(0, p)(0, a)(0, n)(2, m)(2, n)(5, a)(3, a)(0, s)(0, b)(5, a)(1, p)(5, c)(2, k)(0, e)

# top-k LZ78

(0, p)(0, a)(0, n)(2, m)(2, n)(5, a)(3, a)(0, s)(0, b)(5, a)(1, p)(5, c)(2, k)(0, e)

→ **time-dependent** top-k trie references

# top-k LZ78

(0, p)(0, a)(0, n)(2, m)(2, n)(5, a)(3, a)(0, s)(0, b)(5, a)(1, p)(5, c)(2, k)(0, e)

→ **time-dependent** top-k trie references

- hybrid of **LZ78** compression and **Misra-Gries** sketch
- simulate decrements & manage garbage queue using [Metwally et al., 2005]
- constant amortized time per input character

# LZ78 vs. LZ77

Input:  $a^n = aaaaaaaaaaaaaaaa \dots$

LZ78:

LZ77:

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LZ78:  $(0, a) (1, a) (2, a) (3, a) \dots \rightarrow \Theta(\sqrt{n})$  phrases

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# LZ78 vs. LZ77

Input:  $a^n = aaaaaaaaaaaaaaa \dots$

LZ78:  $(0, a) (1, a) (2, a) (3, a) \dots \rightarrow \Theta(\sqrt{n})$  phrases

⊕ simple algorithm (compressed space by default)

⊕ straightforward efficient encoding

LZ77:  $a(1, n-1) \rightarrow 2$  phrases

⊖ much harder to compute

⊖ harder to encode efficiently

→ best of both worlds?

# top-k LZ77

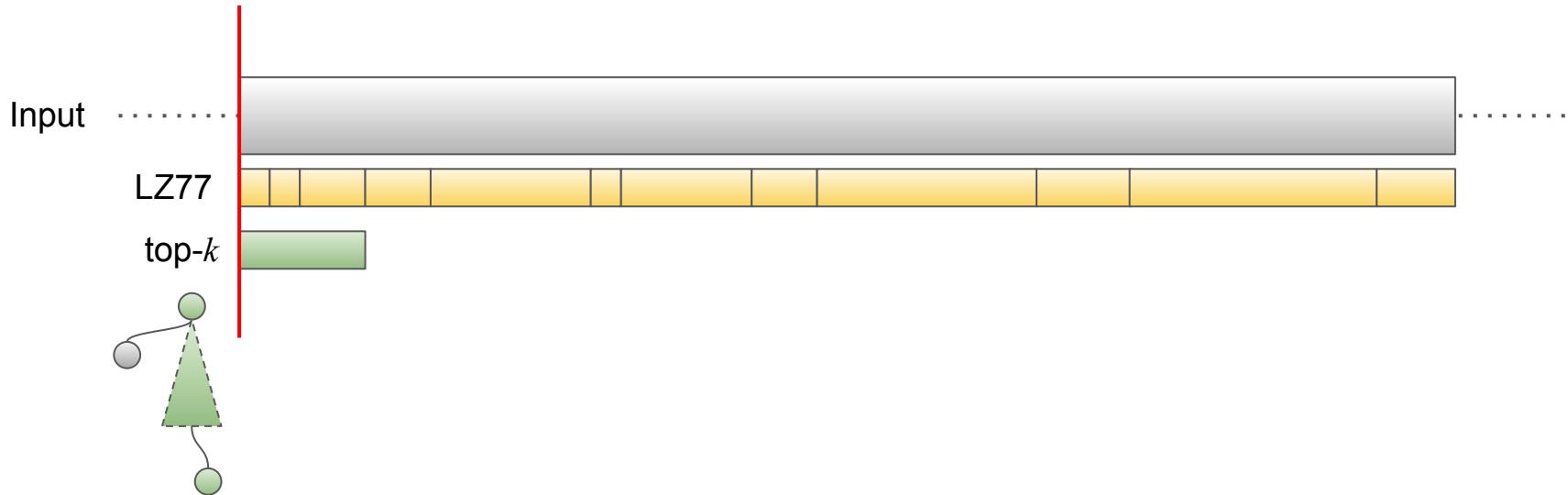
## Algorithm Sketch:

1. Maintain a top-k trie of phrases (like in top-k LZ78)
2. Partition the input into blocks of size  $B = O(k)$
3. Compute the block's LZ77 parsing in time and space  $O(B) = O(k)$
4. Greedily pick the next LZ77 phrase or matching pattern from the trie

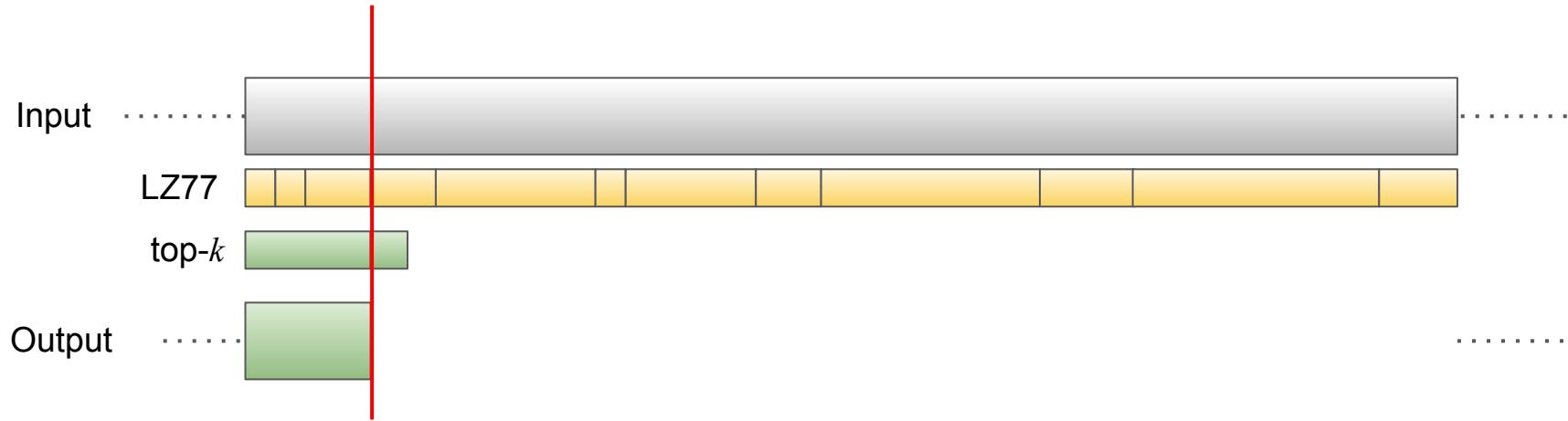
# top-k LZ77



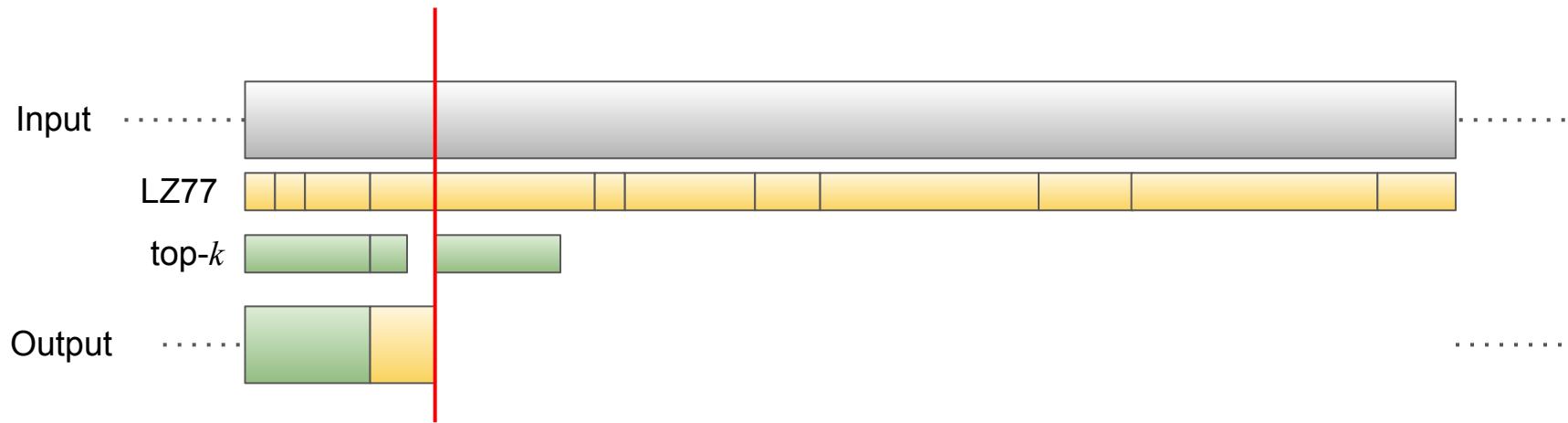
# top-k LZ77



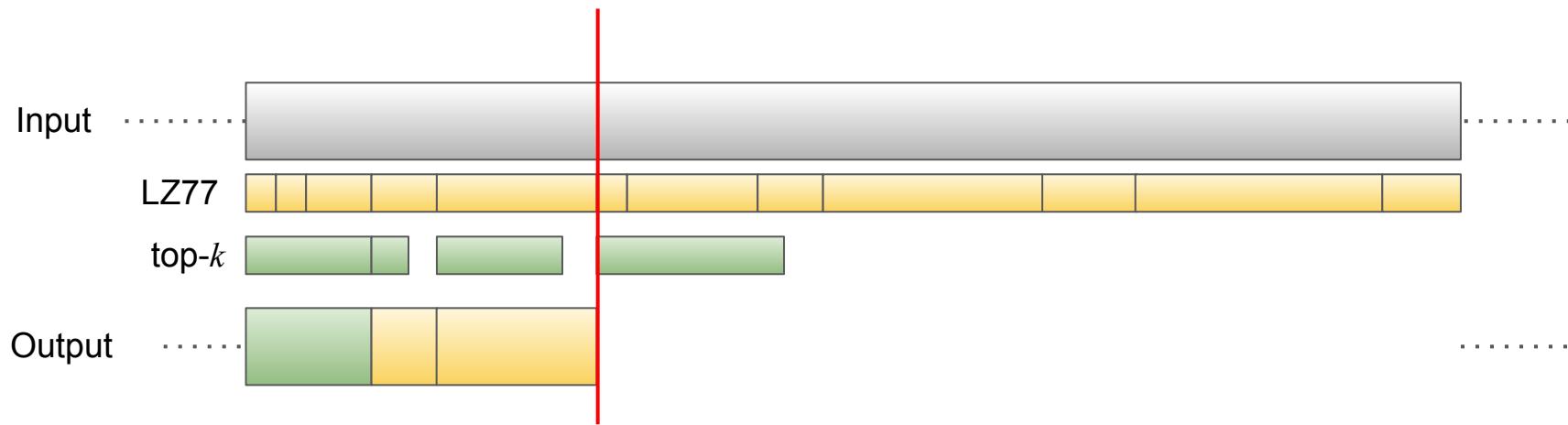
# top-k LZ77



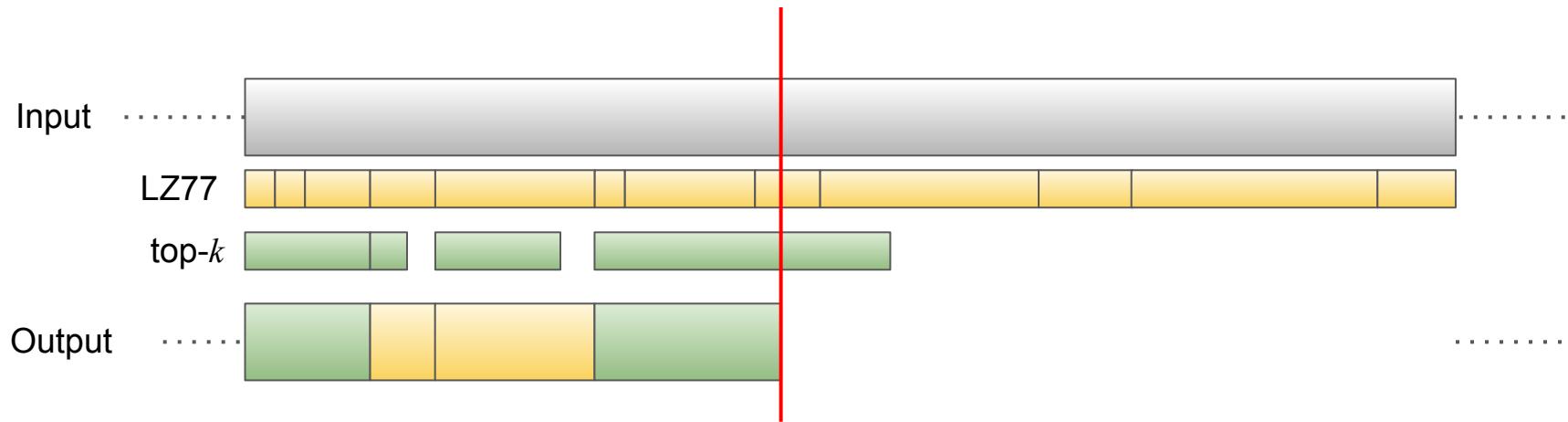
# top-k LZ77



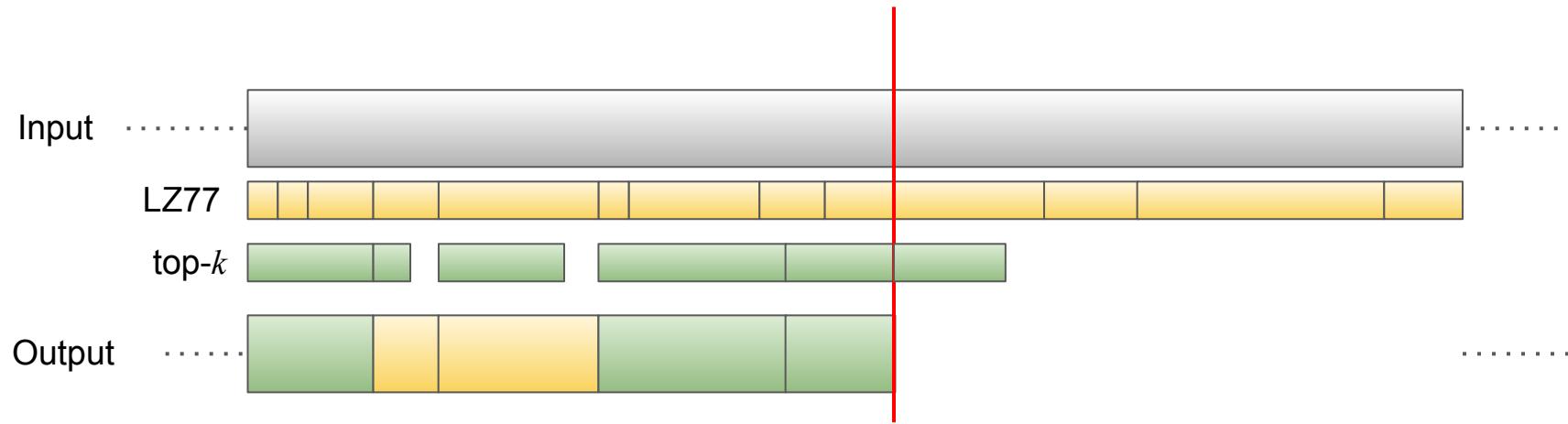
# top-k LZ77



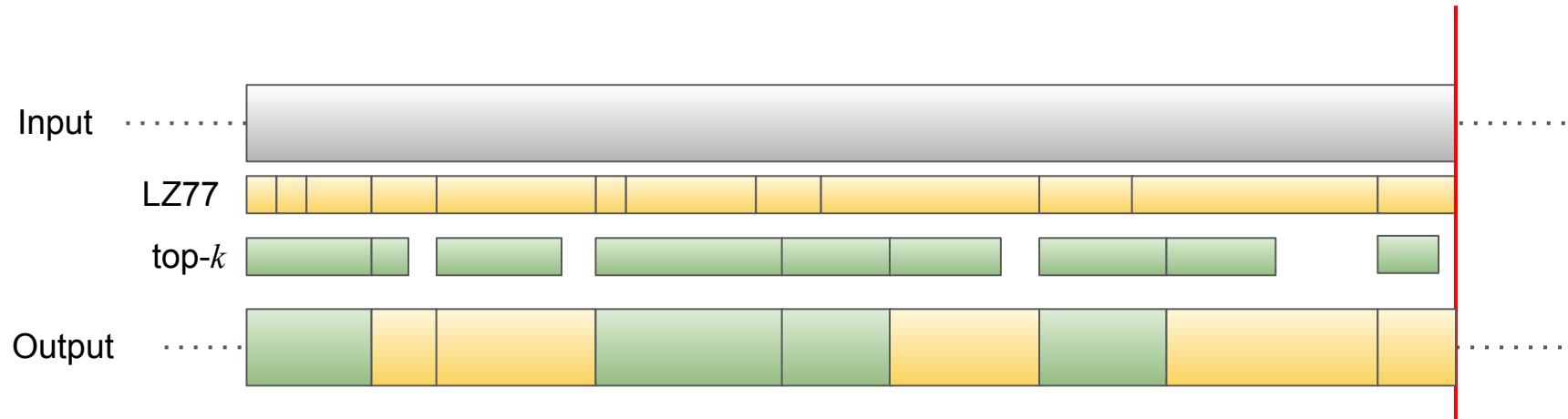
# top-k LZ77



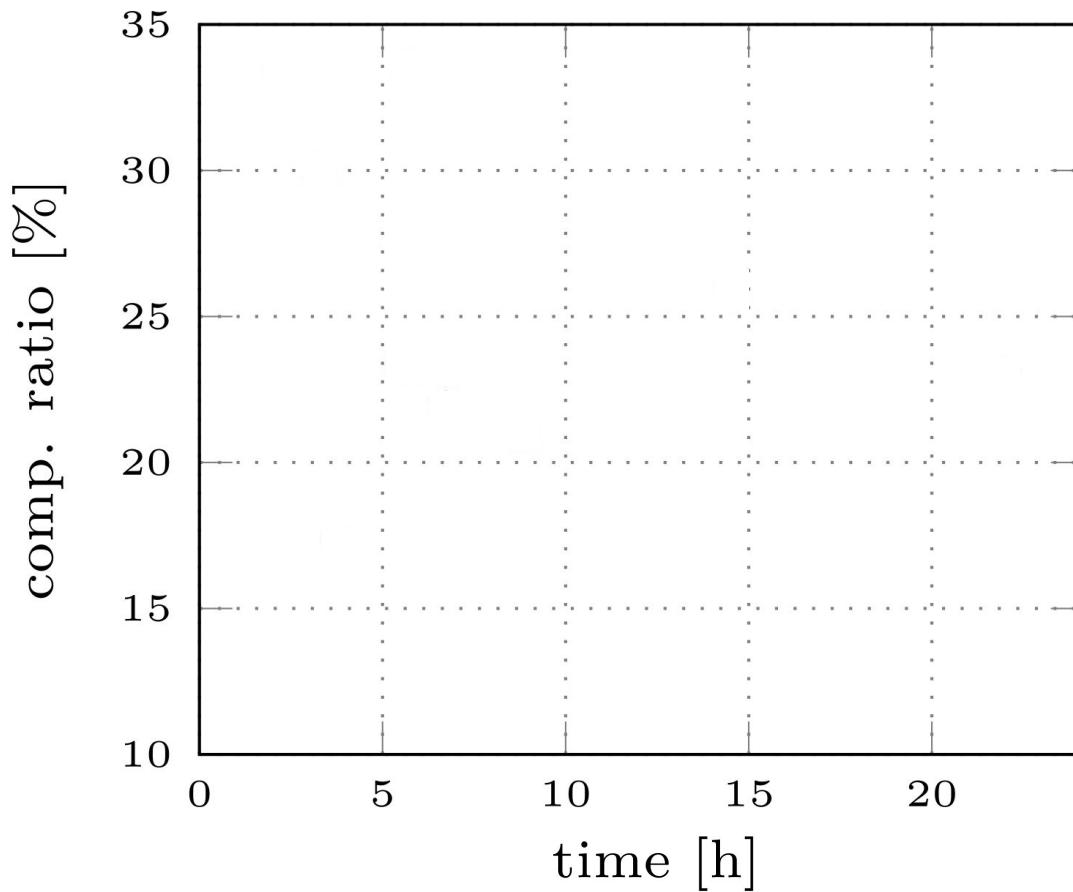
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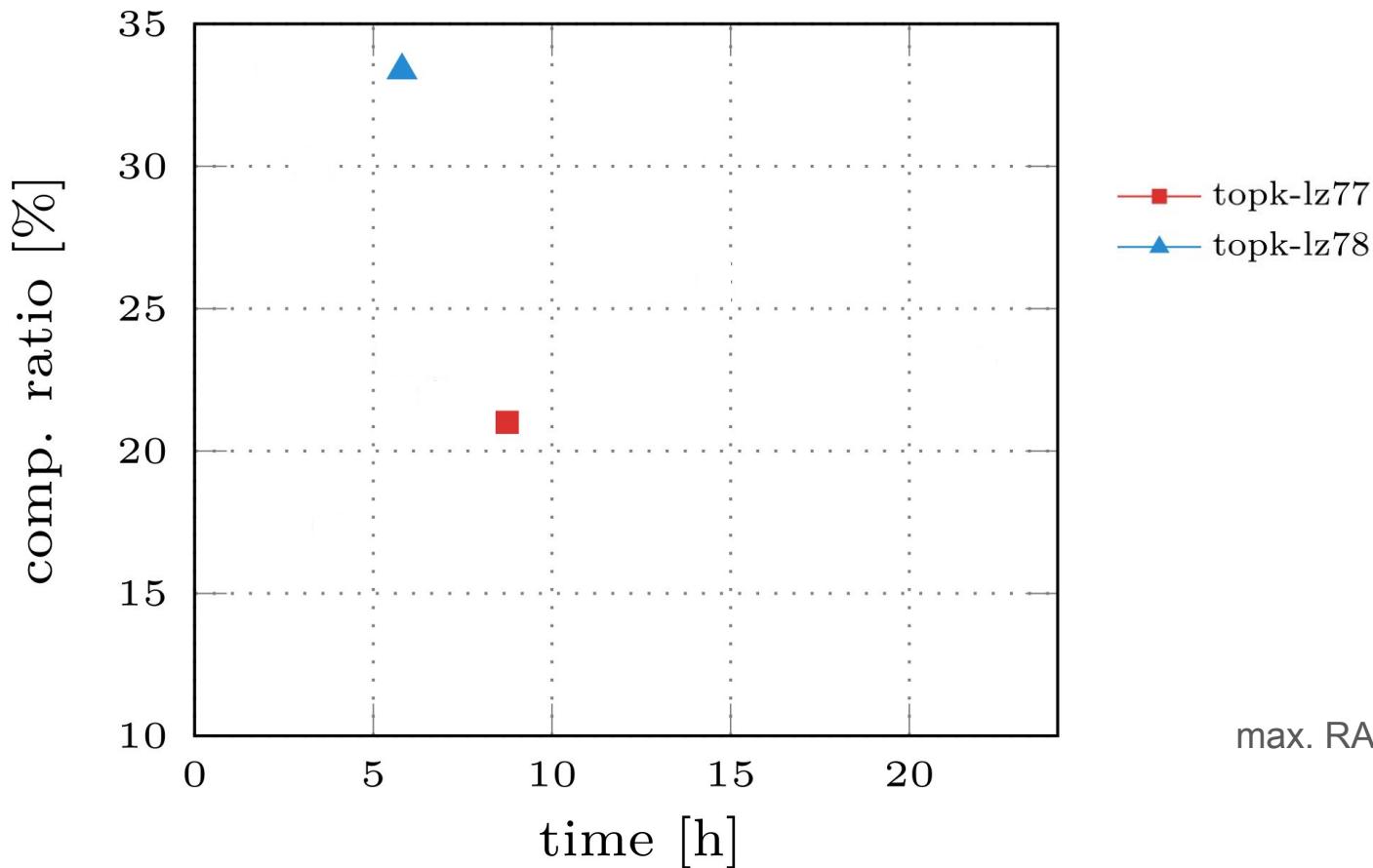


# COMMONCRAWL (100 GiB web text crawl)

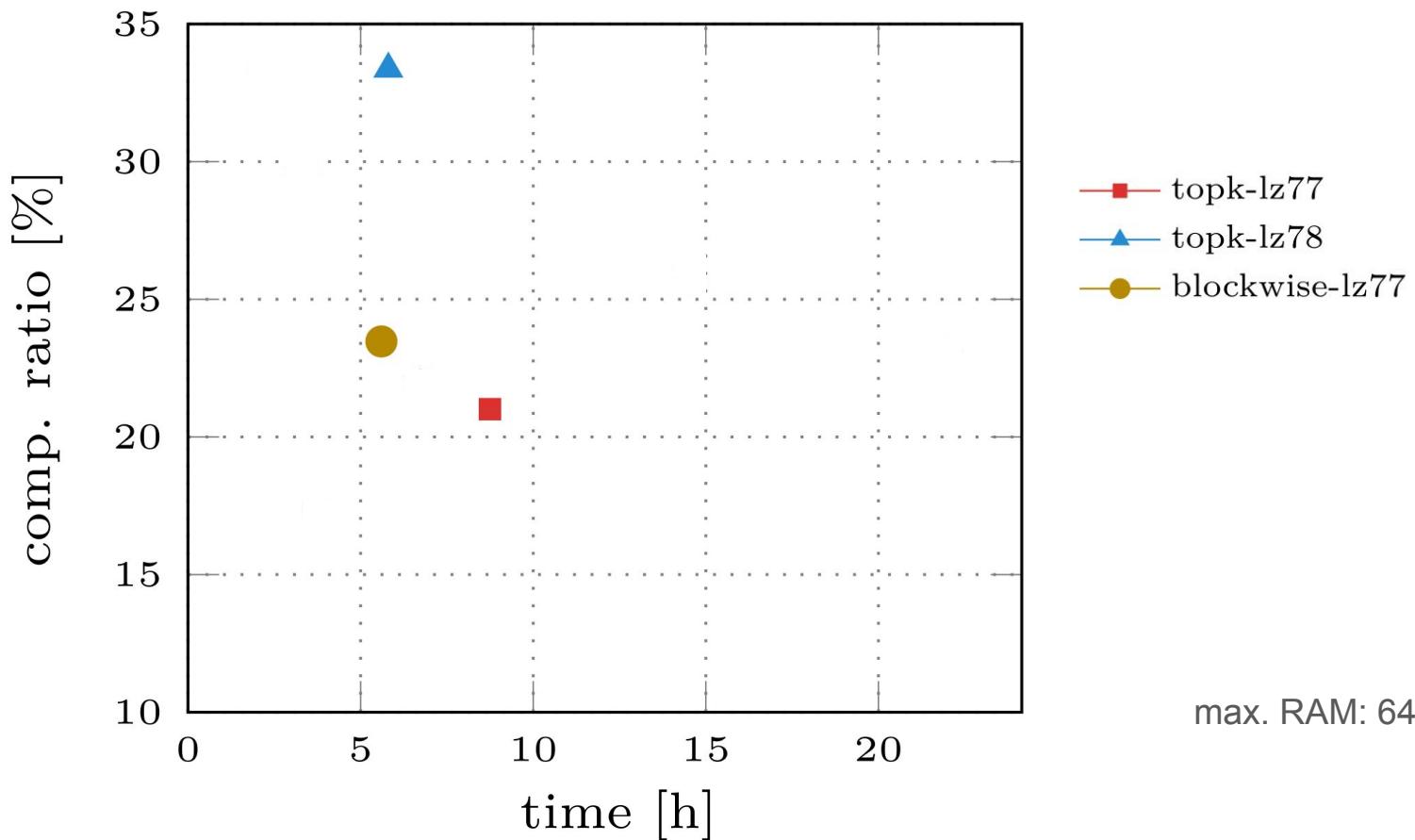


max. RAM: 64 GB

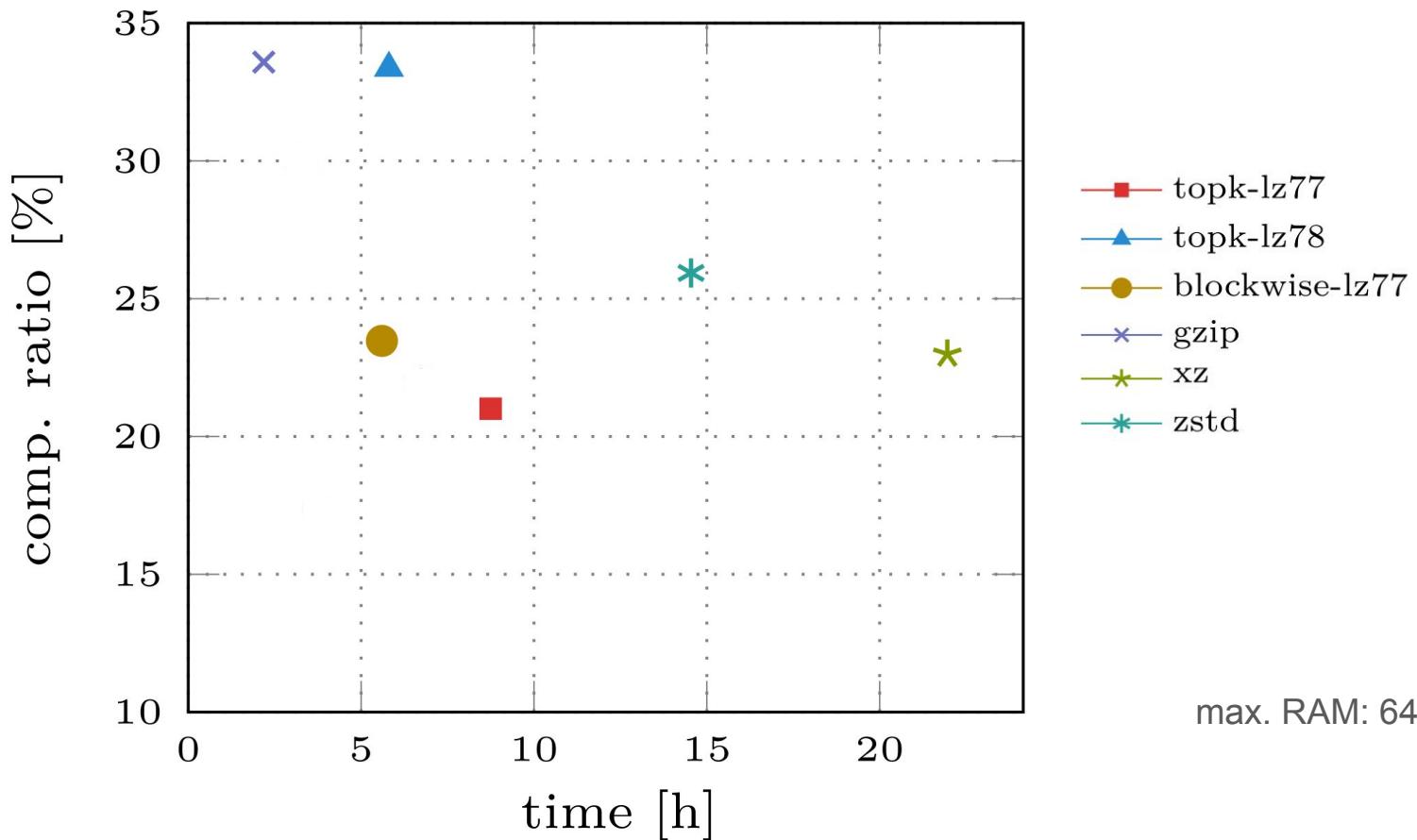
# COMMONCRAWL (100 GiB web text crawl)



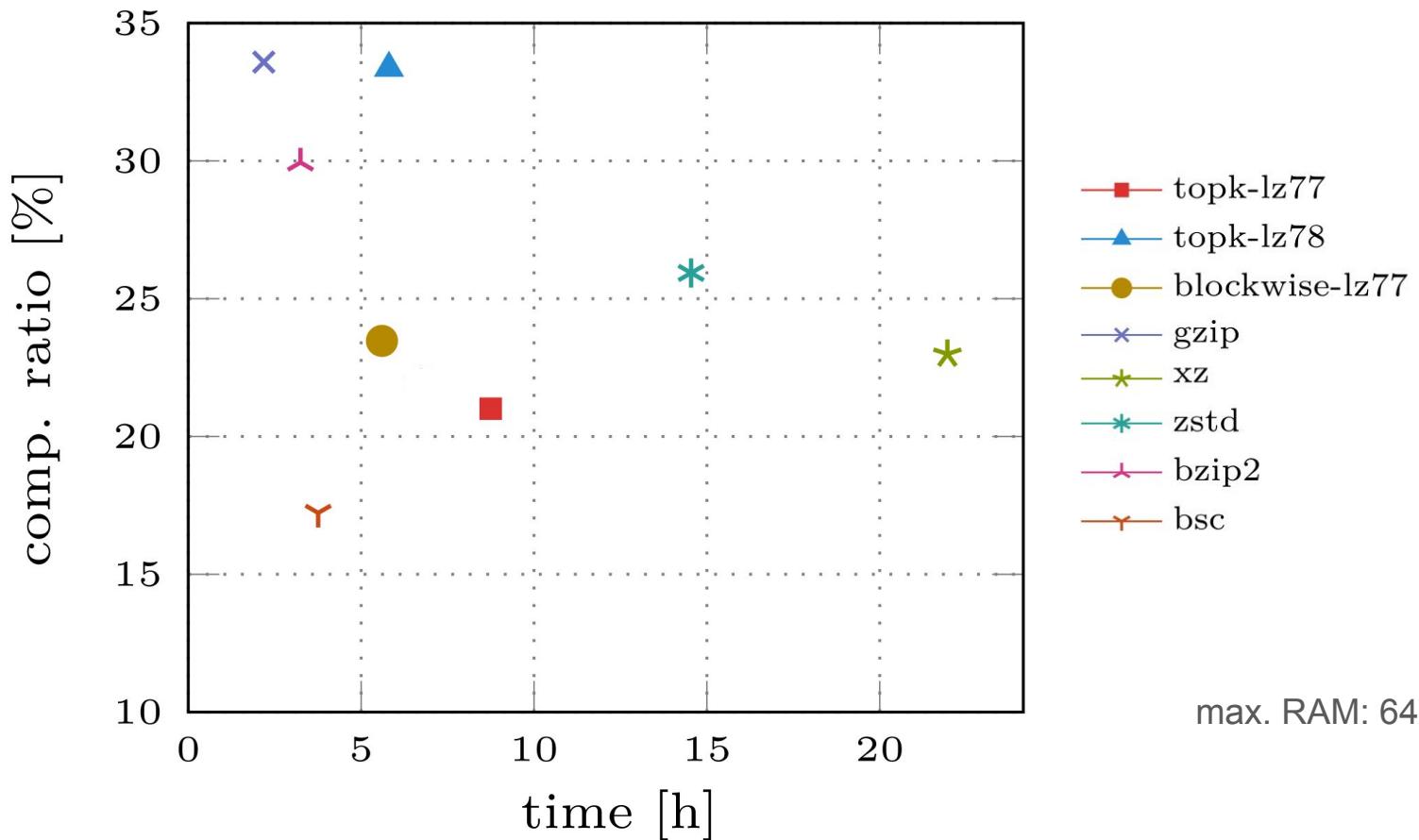
# COMMONCRAWL (100 GiB web text crawl)



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# COMMONCRAWL (100 GiB web text crawl)



# Future Work

- Parallel computation
- Statistical encoding of trie references
- Dynamic string attractors? (as opp. to tries)
- **Precompression** of long repetitions
- Random Access (e.g., like [Arz & Fischer, 2018])
- Information Retrieval
- Your idea here!